Internal migration in a developing country: a panel data analysis of Ecuador (1982-2010)

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Abstract

In this paper, we examine determinants of internal migration flows between the 21 provinces of Ecuador from 1982 to 2010. Using specifications based on the gravity model, we identified push and pull factors. We considered multilateral resistance to migration by using various monadic and dyadic fixed effects structures. The study confirmed the concentration of the population in the two provinces that contain the country’s main cities. However, in recent years, this trend has weakened, to the extent that the provinces with the greatest influx of migrants are not necessarily the most populated. This indicates that growth has become more balanced throughout the territory, and that small and medium-sized cities are increasingly important.

Keywords: internal migration, Ecuador, urban development, gravity model

JEL code: J61, J62, O15

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1. Introduction

International migration has been the focus of much of the media, and even academic, attention for many years. However, internal migration continues to be enormously important, due to its volume and its impact on the configuration of countries (World Bank, 2009): 51% of the world population lived in urban areas in 2010. In that year, 37% of people living in cities were located in urban areas larger than one million inhabitants, while in 1960 that proportion was 39%. Consequently, urbanisation can be seen not only as a megacity phenomenon but also as a process where small and median cities matter more and more. These changes in the distribution of urban population has been the result of vegetative population growth and migration flows. The former has been clearly decreasing over time: between 1960 and 1970 the world population increased at a 2% annual growth rate, while between 2000 and 2010 such growth rate was just 1.2%. This trend leaves a stronger role of migration flows to shape the distribution of population in space within every country. This paper is focused on the analysis of internal migration flows and if and how urbanisation has changed its role as push and pull factor for population moves.

The size and intensity of migration flows depend on circumstances in the place of origin, which could be push factors, and those at the destination, which are pull factors. Migrants subjectively evaluate economic, psychological and social reasons for moving (Todaro, 1980). Faggian et al. (2015) review regional science contributions on interregional migration determinants. One of the key aspects is its role as automatic stabilizer of utility over space. Nevertheless, permanent differentials hold in the long term, due to place specific aspects, including climatic conditions and natural and social endowments or simply due to the considerable stability of variables such as housing provision, which contributes to reducing migration flows and the rate of convergence. There is wide evidence of migration responding to utility differentials (Biagi et al. 2011, Etzo, 2011, Hunt, 2006) and also responding to natural amenities – place specific factors (Partridge et al., 2008, Faggian et al., 2012).

As Barro and Sala-i-Martin (1992) argue, the expected consequence of labour flows is territorial convergence, to the extent that differences that have arisen in income and employment opportunities are tempered, and the initial equilibrium is restored. If salaries and the marginal product of capital are inversely related, population flows are accompanied by capital flows, which accelerate the process. Such economic convergence can take place with or without territorial concentration of economic activity. As stressed in the 2009 World Development Report, territorial concentration and urban agglomeration matters: “an important insight of the agglomeration literature – that human capital earns higher returns where it is plentiful – has been ignored by the literature of labour migration” (World Bank, 2009, p. 158).
At the same time, though, several OECD reports (2009a, b, c) have found that growth opportunities are both significant in big and small urban areas. Following (Barca et al., 2012) “mega-urban regions are not the only possible growth pattern […] context and institutions do matter when we consider economic geography”. Finally, as Duranton and Puga (2000) argue, what matters is the efficiency of the overall “system of cities” and “there appears to be a need for both large and diversified cities and smaller and more specialised cities”.

This debate is key to the design of all economic and social policies. It is also essential to know the causes and conditions that influence migration decisions, in order to understand their nature and anticipate the consequences in terms of economic progress. The size and intensity of migration flows depend on circumstances in the place of origin, which could be push factors, and those at the destination, which are pull factors. Migration contributes to the increase urbanization while making cities much more diverse places (IOM, 2015).

In this short introduction, we focused the attention on world trends and global policy discussions. This paper, though, is focused on internal migrations in a single country, Ecuador. Our work contributes to the empirical literature on interregional migration by analysing a small open developing country. Besides, Ecuador is a country with some circumstances that make it an interesting case study. Let’s start by looking at one of the more urbanised areas in the world: South America. Out of the 22 world subregions1, South America represents the part of the developing world where urbanisation is higher (84% in 2010, only below Northern Europe and Australia & New Zealand, two developed world subregions). As in many other places in the world, the weight of large cities over total urbanization has decreased, from 47% in 1960 to 45% in 2010. Indeed, in 2010 46% of all inhabitants lived in urban areas with small or median cities (below one million inhabitants). Ecuador is one of the countries with lower urbanisation rates in South America, only above Guyana, Paraguay and Bolivia. Even though this figure has risen very fast in the last 50 years, this speed has been lower in larger cities: the proportion of urban population in cities above one million inhabitants represented 52% of total Ecuadorean urban population in 1960. In 2010 this proportion was about 47%. Consequently, Ecuador represents a representative case in which the process of urbanisation is taking place with a strong emphasis on small and medium cities.2

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1 The classification of geographical regions corresponds to the United Nations Geoscheme, which can be accessed at [http://unstats.un.org/unsd/methods/m49/m49.htm](http://unstats.un.org/unsd/methods/m49/m49.htm).

2 Among the other countries with low (below 80%) urbanisation rates in South America, only Paraguay and Ecuador experienced a decrease of the importance of the largest cities, while in Perú, Colombia and specially in Bolivia, largest cities have grown substantially more than the other cities. Both Guyana and Surinam have no cities above one million inhabitants.
Still, persistent territorial inequalities exist. Despite being a small country of around 16 million inhabitants, it has two main cities: Guayaquil and Quito. As the rate of urbanization stood considerably lower than that of neighbouring countries, the process of internal migration is ongoing, and the proportion of the population that lives in cities could be expected to rise in the future. Finally, around 10% of the population of Ecuador has emigrated, which has clearly had a considerable impact on the changes in internal population flows.

We adopted a random utility maximization (RUM) theoretical model, based on differences in economic expectations between the provinces of origin and destination. We used census data from 1974 to 2010 to propose and estimate a model of interregional migration, in which we analysed various key factors: population, distance, production structure and urbanization. We also controlled for factors that affect the selectivity of migration, such as age structure or level of education. Finally, we calculated a set of models, including a series of monadic and dyadic fixed effects that enabled us to control various expressions of multilateral resistance to migration. In addition to confirming the importance of a sector’s structure on migration flows, we observed that population flows were to the most populated provinces, but the pace of concentration had dropped gradually over time. If this trend is confirmed, we can state that there is a process of territorial balance in Ecuador, in which the growth of cities in provinces is balancing the territory.

The next section of the document presents the case study of migration between Ecuadorian provinces. Section 3 describes the theoretical framework of reference, and defines the empirical specification. The results of the calculations are given in Section 4, and Section 5 contains an analysis of the sensitivity and robustness of the calculations. The paper ends with the main conclusions of the study, and some policy recommendations.

2. Internal migration in Ecuador

Economic and social context in Ecuador

Ecuador is the eighth largest economy in Latin America (LA). The country is divided into 24 provinces\(^3\), grouped into four geographical macro-regions: the Coast, the Andes, the Amazon and Islands (Graph 1). The average growth in GDP in the last 50 years was 4%, and there have been a combination of deep recessions (the last one was in 1999), and strong periods of expansion (12% a year between 1973 and

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\(^3\) Ecuador has 24 provinces, 221 cantons and 1149 parishes. While we acknowledge that more detailed data could improve our results, for instance looking at more recent census which have information of migration flows between cantons, we opted to work with province data in order to enlarge the analysed periods.
Since 2000, the country has grown at a rate of 4.5% per year (World Bank, 2016). In addition to oil revenue, remittances from emigrants represented 2.3% of GDP in 2014; an amount similar to the total revenue from non-oil exports (CEPAL, 2005). Table 1 provides an international comparison of some of the social and economic indicators.

**Table 1. International comparison of Ecuador’s social and economic indicators**

4 The System of Social Indicators of Ecuador (SIISE, 2014) considers that a person is in extreme poverty if they meet two or more of the following conditions: 1. Their dwelling has inadequate physical characteristics; 2. Their dwelling has inadequate sanitary systems; 3. The household has high economic dependence; 4. There is a child (or children) in the household who does not go to school; 5. The dwelling is in a critical state of overcrowding.

5 The poorest tenth of the population received 1.86% and 1.85% of the national income between 1988 and 2012, whilst the richest tenth received 34.1% and 33.7% in the same period.
| Indicators                                      | Ecuador | Brasil | Colombia | México   | Perú    |
|------------------------------------------------|---------|--------|----------|----------|---------|
| GDP per capita (2014, PPP U.S. $)              | 6,346   | 11,384 | 7,904    | 10,326   | 6,541   |
| Gini index (2013) %                            | 43.7    | 52.9   | 53.5     | 48.1     | 44.7    |
| Unemployment rate % (2014)                      | 4.6     | 6.8    | 10.1     | 4.9      | 4.2     |
| Urban populations % (2014)                      | 64      | 85     | 76       | 79       | 78      |
| Migration recent interregional * (2001)         | 5,24    | 3,39   | 8,1 †    | 4,4 ††   | 8,6 †   |

Sources: World Bank (2016) and * Rodríguez Vignoli, (2004). † Calls for 1993 data and †† for 2000.

There is considerable disparity in Ecuador at regional level. Four provinces are home to 62% of the total population and concentrate 70% of the economic activity (INEC, 2014; BCE, 2014): Pichincha (which is where the capital, Quito, is situated) and Azuay in the Andes; Guayas (whose capital is Guayaquil, the largest city in Ecuador) and Manabí on the Coast. The poorest provinces are Bolívar (the Andes), Los Ríos and Esmeraldas (the Coast), and Napo (the Amazon). In 2010, Pichincha and Guayas were specialized in the service sector, whilst in small and medium-sized provinces, such as Cañar, Cotopaxi, Chimborazo, Napo and Sucumbíos, the main employment (almost 50% of the population) was in agriculture. The secondary sector was strongest in the provinces of Azuay, Imbabura, Loja and Zamora Chinchipe.

**Distribution of the population and internal migration in Ecuador**

Between 1950 and 2010, the population of Ecuador increased fivefold: the urban population expanded tenfold, the rural population twofold, and the rate of urbanization rose from 29% to 63%. In 2010, the most urbanized provinces were the Galápagos, Guayas, El Oro and Pichincha. Between 1982 and 2010, Pichincha and Guayas gained importance: they housed 42% of the population between them in 1982, and 48% in 2010. This indicates that a process of urban concentration has taken place at the same time as the urbanization process in the country. However, more recently, the rate of growth of these two provinces has slowed down noticeably, while the population grew above the average rate in some Amazonian provinces: the 11 provinces with the lowest population (mainly Amazon) together housed 12.4% of the total population in 1982, compared to 12.1% in 2010. In other words, these provinces have hardly decreased in relative importance. In fact, between 2001 and 2010, the relative importance of this group of provinces increased compared to the national situation overall. The drop in the rate of concentration of the population is notable and demonstrated by the lower migration rates in 2010, and in the lower rate of growth in Pichincha and Guayas. Consequently, “the most populated provinces are not necessarily those that grow most” INEC (2012).
The changes in the relative importance of each province can be explained partly by transformations in the country’s production structure. A major downturn in the manufacture and exportation of Panama hats in the 1950’s led to migration towards the rural areas of the Coast, the Amazon, and abroad (Espinoza and Achiag, 1981). In the 1960’s there was another major migration process, due to changes in the agricultural export model (Pachano, 1988). The oil boom (the first oilfield was found in 1962) and the “process of colonization” made the Amazon a new destination for migration (Guerrero and Sosa, 1996).

In the 1980’s Ecuador was affected by fluctuations in oil production and exportation, natural disasters and the military conflict with Peru. In 1999, the Ecuadorian economy suffered a serious economic and financial crisis that had severe effects on unemployment and poverty. It led to high emigration to other countries, particularly Spain (Bertoli et al., 2011). In 2000, Ecuador introduced dollarization, and a period of economic stability began. Internal migration between provinces tended to drop, and many emigrants began to return, due to the international recession and backed by government policy.

Against this background, we analysed changes in interprovincial migration in Ecuador, using census data from 1982 onwards. Censuses can be used to calculate migration rates by comparing the province of residence with the province of birth (permanent migration or stock), and with the province of residence five years before each census (recent migration or flow). Whilst the migrant stock has gradually increased, from 18.5% in 1982 to 20% in 2010, the flow (recent migrants) has fallen from 8.3% in 1982 to 4.7% in 2010. Therefore, there is a decreasing trend in internal migration flows. Consequently we do not analyse international migration flows.

The same decreasing trend can be found in Latin America (CEPAL, 2007) for various reasons, according to Rodriguez Vignoli (2004): the replacement of internal migration by international migration; the increase in daily journeys for work or study, which eliminate the need to migrate; the increase in home ownership associated with rising incomes; and a slow-down in migration flows from the countryside to the city due to the expansion of urbanization. Rodriguez Vignoli does not consider that this is due to a process of regional convergence. According to CEPAL (2012), in Ecuador migration between areas

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6 Since the twenty-first century, measures have been in place to promote the colonization of these areas of the country. In 1885, the “Eastern Province Act” was brought into force, to encourage settlement in the East and to control borders, as Peru was expanding due to rubber activity, which was booming at that time. Among other matters, this Act approved the granting of financial incentives and the free allocation of plots of land to people who moved to the East, as well as various financial benefits for growers of rubber, Chincona, coffee and cacao (Esvertit, 2005). At the start of the 1960s, an agreement was reached to recolonize and resettle the East, to stimulate the impoverished agricultural sector. At the end of 1959, the government obtained funding from international organizations to support this project. In 1964, the Agricultural and Settlement Act was approved, and the Ecuadorian Institute of Agrarian Reform and Settlement (Instituto Ecuatoriano de Reforma Agraria, IERAC) was created to implement the new legislation. The programme involved actions on state properties, followed by semi-public and private properties for social purposes, and finally private properties, and generally followed the FAO’s recommendations (González, 1983).
continues and is associated with the multipolar economic development of the country, and the persistence of chronic poverty in some provinces, which push the population mainly to dynamic provinces or those with greater opportunities and resources.

Appendix 1 shows the percentage of net migration and the proportion of the total population in each province. Between 1982 and 2010, except in Pichincha (where the metropolitan area of the capital, Quito, is situated), all of the Andean provinces were affected by out-migration. Nevertheless, from the 2001 census onwards, this trend was reversed in some Andean provinces, which became net recipients of migration. In the Coast region, the provinces of Guayas and El Oro, which had been net recipients of migration, became less attractive to migrants, while the rest of the provinces in the Coast region have net emigration. The eastern provinces (in the Amazon region) have become less attractive to migrants overall. As a result, from 2001 most Amazonian provinces were affected by out-migration, although two of them (Pastaza and Orellana) remained attractive to migrants (expansion of the demographic border and mining, CEPAL, 2012). In the four censuses that were analysed, the Galápagos attracted migrants, despite the urban development regulations and the laws on residence in the archipelago, which are designed to protect the ecosystems and biodiversity. In fact, since 1990 approximately two-thirds of the resident population in the Galápagos was born outside the province.

3. The gravity model of internal migration in Ecuador, 1982-2010

Theoretical framework and empirical specification

According to models designed by Lewis (1954), Todaro (1969, 1980) and Harris and Todaro (1983), migrants move from rural or undeveloped areas, with high unemployment and underemployment rates, poor working conditions and low salaries, to developed, urban areas, with higher levels and/or rates of productivity growth, as well as better education opportunities, health care and quality of life in general (Royuela et al., 2010). Any place can be considered a centre of production and consumption, although urban centres are at an advantage, as they benefit from positive externalities (agglomeration economies), although excessive concentrations could lead to problems of congestion and social inequalities (Henderson, 2003). The neoclassical theory of migration is based on the concept of utility maximization: after a cost-benefit analysis, each individual decides whether or not to migrate, and to which destination (Borjas, 1988 and 1999). The literature also assumes that migration is selective and depends on individual characteristics, including sex, age and level of education.

From an aggregate perspective, and taking the work of Ravenstein (1885) as a starting point, migration models have drawn heavily on gravity models. The economics literature has developed models that result
in gravity specifications. Our study is based on a theoretical development given in Beine et al. (2015). Thus, migration from the region of origin \( j \) to the region of destination \( k \) in the period \( t \) \((m_{jkt})\) is a function of the proportion of people who migrate \((p_{jkt})\) and the stock of population living in \( j \) \((s_{jt})\).

\[
m_{jkt} = p_{jkt}s_{jt}
\]  

This is the starting point for the RUM model, which assumes that the utility \( U_{ijkt} \) of an individual \( i \) moving from \( j \) to \( k \) at a time \( t \) depends on \( w_{jkt} \), the deterministic utility gained by individual \( i \) due to moving from \( j \) to \( k \) in \( t \); \( c_{jkt} \), the costs of moving from \( j \) to \( k \) in time \( t \); and \( \varepsilon_{ijkt} \), an individual stochastic component of utility:

\[
U_{ijkt} = w_{jkt} - c_{jkt} + \varepsilon_{ijkt}
\]  

The assumptions about the distribution of the stochastic term in Equation (2) determine the expected probability of selecting destination \( k \). If it is assumed that \( \varepsilon_{ijkt} \) is stochastic, independent and identically distributed, according to extreme value type 1, and that the deterministic component of utility does not vary with the origin \( j \) (the expected average utility of not migrating is normalized to zero), then the expected gross migration flows from \( j \) to \( k \) could be close to gravity equation (3):

\[
E(m_{jkt}) = \phi_{jkt} \frac{\gamma_{kt}}{\Omega_{jt}} s_{jt}
\]  

Where: \( \gamma_{kt} = e^{w_{jkt}} \), \( \phi_{jkt} = e^{-c_{jkt}} \), \( \Omega_{jt} = \sum_{t \in D} \phi_{jlt} \gamma_{lt} \).

According to Equation (3), expected migration flows depend multiplicatively on (i) \( s_{jt} \), which is the capacity of expulsion from \( j \) in \( t \); (ii) \( \gamma_{kt} \), the capacity of attraction of the destination region \( k \); and (iii) \( \phi_{jkt} < 1 \), the accessibility of the destination region \( k \) to potential migrants from \( j \). Expected migration flows are inversely related to (iv) \( \Omega_{jt} \), which represents the expected utility of potential migrants from the situation of origin. The value of this last element increases when accessibility rises \((\partial \Omega_{jt}/\partial \phi_{jlt} > 0)\).
0), which means that enhanced accessibility of an alternative destination \( l \) will invariably lead to a drop in the expected bilateral flow of migration from \( j \) to \( k \).

The property of independence of irrelevant alternatives (IIA), which is derived from a distribution according to McFadden (1974) of the stochastic component of utility in (2), implies that a variation in the attractiveness or accessibility of an alternative destination \( (l) \) leads to a proportional, identical change in \( E(m_{jkt}) \) and \( E(m_{jjt}) \). To move from terms of mathematical expectation to an expression based on data, we must add to Equation (3) a component of the error term \( \eta_{jkt} \), with \( E(\eta_{jkt}) = 1 \), to obtain the classic gravity model in the literature on migration:

\[
m_{jkt} = \phi_{jkt} \frac{Y_{kt}}{\Omega_{jt}} s_{jt} \eta_{jkt}
\]

(4)

The IIA axiom may not be true for various reasons, which lead to that known as multilateral resistance to migration. According to Bertoli et al. (2011), the scale of migration flows between two destinations depends not only on their relative attractiveness, but also on the attractiveness of alternative destinations. Therefore, an increase in the attractiveness of a third destination will decrease the probability of migration flows between the two initial destinations. If this concept is overlooked, biased estimates could be produced (Bertoli et al., 2013b). Multilateral resistance to migration may arise when assumptions about the distribution of the stochastic component are altered, or if we consider the sequential nature of migration decisions.

Population groups in the place of origin may be heterogeneous, and as a result the same destination may have a different level of attractiveness for them, for example, due to sex, age, level of education or aspects associated with the psychological costs for different population groups. The existence of this heterogeneity introduces a pattern of correlation with all destinations into the stochastic component of utility. According to Bertoli et al. (2013a), if a correlation is assumed to exist in the stochastic component of utility, then an increase in the attractiveness of a third destination that is perceived as a substitute for \( k \) will reduce the volume of migrants between \( j \) and \( k \) \((m_{jkt})\) proportionally more than the volume of individuals who decide to remain in the place of origin \((m_{jjt})\).

Similarly, we can assume that the model should include not only the present characteristics of the alternative destinations, but also the future expectations of each one of them (in \( t+1 \)). Even if we assume that the stochastic component of utility is independent and identically distributed (IID) and of extreme
value type 1, the final model will be sensitive to expectations about the future attractiveness of alternative
destinations (Bertoli at al., 2013b, Beine and Coulombe, 2014). Therefore, in accordance with Hanson
(2010) and Beine et al. (2015), traditional models explain migration flows as a result of different
characteristics in the place of origin and destination, assuming the IIA property and therefore avoiding
multilateral resistance to migration. However, the impact of conditions in the place of origin tends to be
overestimated if the influence of alternative destinations is not considered.

These effects have been controlled in various ways in the literature. When the panel is large enough in
terms of cross-sectional and longitudinal data, the multilateral resistance term adapts to the structure of
the common correlated effects estimator (CCE, Pesaran, 2006) as used, for example, in Bertoli et al.
(2013b) and Bertoli and Fernández-Huertas Moraga (2013). Our case study does not have the right
characteristics to apply this type of techniques. Consequently, we follow the applied migration literature
(see below) and we aim to capture these aspects by using various dummy variable structures, as we have
three data dimensions (origin, j, destination k, and moment in time t). Therefore, in this study, we
estimate models with different fixed effects structures, from the simplest model which is a priori biased,
to more complex structures that lose part of the information, but enable us to estimate parameters that
are free of some biases.

1. Basic model with origin and destination variables and time fixed effects.

\[
\ln m_{jkt} = \alpha + D_t + \beta_1 X_{jkt} + e_{jkt} \tag{5}
\]

Where: \( \alpha \) is the intercept, \( D_t \) is the vector of dichotomous variables for each year, and \( X_{jkt} \) is the vector
of independent variables in the model. The vector of dummy variables for each year enables us to control
common disturbances in time in all provinces. However, if multilateral resistance to migration exists,
this model will produce biased estimates. The inclusion of time fixed effects enables us to capture general
time shocks in all observations. In turn, this enables us to capture the multilateral resistance to migration
of potential destinations that are not included in the database.

2. Panel model with monadic fixed effects of the origin and the destination, and time fixed effects.

\[
\ln m_{jkt} = D_j + D_k + D_t + \beta_1 X_{jkt} + e_{jkt} \tag{6}
\]

8 See Beine et al. (2011), McKenzie et al. (2013) and Ortega and Peri (2013) for different justifications for the
inclusion of these dummies.
Where \( D_j \) and \( D_k \) correspond to dichotomous variables for each origin and destination province respectively, \( X_{jkt} \) is the vector of independent variables in the model, and \( D_t \) is again the vector of time fixed effects. Mayda (2010) includes fixed effects of the origin and destination, to control for specific effects of each origin / destination that are not captured by deterministic components of utility. This is the traditional strategy for capturing multilateral resistance to migration in cross-sectional studies. Mayda (2010) uses it to control for the effect of migration policy that is common to all spatial units. In our case, it would enable us to capture the permanent migration policy for the Amazonian provinces or the legal restrictions to immigration in the Galápagos. Nevertheless, this model does not account for most types of multilateral resistance described above.

3. **Panel model with dyadic fixed effects of origin-destination.**

\[
\ln m_{jkt} = D_{jk} + D_t + \beta_1 X_{jkt} + e_{jkt} \tag{7}
\]

Where: \( D_{jk} \) is the vector of dichotomous variables of origin-destination. This specification is similar in nature to the above, with the added feature that we can now quantify specific deterministic effects for each pair of regions (Ortega and Peri, 2013). This fixed effects structure captures any specific bilateral relationship between \( j \) and \( k \), which reflect fixed migration costs, geographic aspects and historic migration networks between pairs of regions, and even permanent migration policies. This structure also includes constant specific characteristics of the origin and destination. However, the distance variable is not included in this model: to the extent that it remains constant for each pair of regions (Karemera et al., 2000; Ortega and Peri, 2013), it shows perfect multicollinearity with this fixed effects structure.

4. **Panel model using origin variables of origin and dyadic fixed effects of destination-time.**

\[
\ln m_{jkt} = \alpha + D_{kt} + D_j + \beta_1 X_{jt} + e_{jkt} \tag{8}
\]

Where: \( D_{kt} \) is the vector of the destination’s dyadic dummy variables for each year, while \( X_{jt} \) is the vector of independent variables of the origin regions. This panel model enables us to control for all of the “pull” determinants of migration, and particularly multilateral resistance derived from heterogeneity in the future perspectives of the destination regions (Beine and Parsons, 2012). The structure also enables us to control for time-invariant characteristics of migration policies that are the same for all provinces. Beine and Parsons (2012) used dyadic fixed effects of destination-time to control for any specificity between potential destinations for any period of time. This method can be used to control for bias in the parameters of the origin variables due to multilateral resistance caused by the heterogeneity of expectations about each destination. Like Beine and Parsons (2012), we also include fixed effects for
each origin. To be clear, this will be our preferred model to capture the parameters associated to the characteristics of the origin.

5. Ordinary least squares with destination variables and dyadic fixed effects of origin-time

\[ \ln m_{jkt} = \alpha + D_{jt} + D_k + \beta_1 X_{kt} + e_{jkt} \]  

(9)

Where: \( D_{jt} \) is the vector of the origin’s dichotomous variables for each year. This method enables us to control for all the “push” determinants of migration, as well as the multilateral resistance derived from heterogeneity in migration preferences by origin. Ortega and Peri (2013) use these dyadic fixed effects of origin-time to control for any specificity in the place of origin in any period of time. This approach can be used to eliminate the bias in the parameters of the destination associated with multilateral resistance due to heterogeneity in preferences of migration by origin. Bertoli and Fernández-Huertas Moraga (2012) use a cross-section model to estimate heterogeneity in preferences by destination subgroups, with dummy variables to control for these subgroups, which we do not consider in this study. To be clear, this will be our preferred model to capture the parameters associated to the characteristics of the destination.

Given the nature of the proposed panel of models, a structure of the random term that is only associated with idiosyncratic errors \( e_{jkt} \) could be assumed for the estimation. Alternatively, we could also assume the existence of a structure composed of permanent individual errors, corresponding to the fixed structure of the panel, that is, for each pair of regions \([\xi_{jk}]\). The second case represents a random effects model, which increases the efficiency of the estimation. Given that in most cases the fixed effects structures will control for unwanted consequences of random effects models, we will generally use the panel estimation assuming the existence of specific random effects for each pair of regions.

One issue not covered in the theoretical approach of the model presented here is the selection of deterministic factors for the function of the utility of individuals. In the empirical literature, the most common aspects are pull factors and opportunities to earn an income at the destination (Mayda, 2010), the gap in income per capita between the origin and the destination (Ortega and Peri, 2009), the population in the place of origin and the income at the destinations (Karemera et al., 2000), the differences in terms of quality of life (Faggian and Royuela, 2010) or the level of urbanization (Royuela, 2015). Given the assumption of normalization in the utility at origin, the deterministic component of utility in the empirical model measures the effect of increasing the gap in expected benefits between the origin and the destination. According to empirical literature on migration, we can apply the relative
difference in the deterministic component of utility to the variable that approximates material welfare (Beine and Parsons, 2012, use the log of the income ratio, while Ortega and Peri, 2009, analyse both linear and logarithmic differentials).

Attractiveness is estimated using the distance between the origin and the destination, which can be determined physically (using the Euclidean distance or distance by road) or economically (the average distance in terms of time), or it can be derived from differences in terms of language, customs, history, culture, and institutions (Belot and Ederveen, 2012; Caragliu et al., 2012).

Very few studies on this research area relate to Ecuador. Studies that do refer to this country mainly focus on the influence of international migration. Notable studies are those by Gratton (2007) on the characteristics of migration from Ecuador to Spain and the United States; Bertoli et al. (2011 and 2013a), who analysed how migration policy redirected traditional migration to the United States to Spain; a study on the relation between migration, remittances and environmental variables in rural communities of Ecuador by Gray (2009 and 2010); and, finally, an analysis of how migration affects family structure and fertility (Laurian et al., 1998).

**The empirical model for Ecuador**

The proposed empirical model analyses the flow of recent migrations by province of origin and destination, using databases of Ecuador’s Census of Population and Housing (CPV) from 1982, 1990, 2001 and 2010. Today, Ecuador has 24 provinces. The provinces of Sucumbíos and Orellana were created in 1989 and 1998 respectively, when they were separated from Napo province. Santo Domingo de los Tsachillas and Santa Elena, which had belonged to the provinces of Pichincha and Guayas respectively, became provinces in 2007. To work with a standardized database for the entire period, we added data from the province of Orellana to that of Napo, Santa Elena to Pichincha, and Santo Domingo to Guayas, to obtain a total of 21 provinces. We did not take into account non-delimited areas, as these are not representative at national level.9 Recent internal migration refers to the population that changed residence in the five years prior to the census. Consequently, it does not include migration that occurred at an earlier time. To avoid problems of simultaneity, the explanatory variables in the model refer to the previous census, so that migration flows between 2005 and 2010 are explained by the characteristics of the provinces in 2001. As the dependent and the explanatory variables are separated five years, we

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9 We did not consider migration data for Sucumbios for 1982, as this information was added to that of the province of Napo. In Ecuador, 1,419 km² of the territory is not assigned to any province. In 2010, these territories had 32,384 inhabitants and corresponded to the areas of Las Golondrinas, La Manga del Cura and El Piedrero.
minimize the chance that a shock could be affecting simultaneously both the dependent and the explanatory variables. As in Rupasingha et al. (2015) and Levine et al. (2000), we understand that future migration does not affect current levels of explanatory variables. We also used information from the 1974 census to calculate control variables for the characteristics of the origin and destination of migration flows for the 1982 census. Finally, as Beine et al. (2015) argue, “controlling for multilateral resistance to migration can make instrumentation unnecessary as long as the endogeneity problem is not due to reverse causality, or as long as the resistance terms capture a big part of the omitted factors” (p.9).

As explanatory variables, we included basic gravity factors: the population of origin and destination and the distance between them, which approximates the costs associated with migration (Peeters, 2012). The distance variable can be measured in kilometres and expressed in logarithms ($L\text{ Dist}$), as in Mayda (2010), or in terms of time ($L\text{ Time}$), which is closer to the economic concept of the cost of moving$^{10}$. We used proportions of employment by branches of activity to control for the sector structure. In general, greater importance of the agricultural sector is traditionally associated with a lower level of development. The most developed provinces were expected to have a higher proportion of people employed in manufacture ($\text{Manufacturing Ind}$) and services ($\text{Services}$) and less employment in primary activities, including agriculture ($\text{Agriculture}$) and mining and quarrying ($\text{Mines & Quarrying}$). The construction sector was also included in the analysis ($\text{Construction}$). We also considered the characteristics of the labour market ($\text{LM}$) according to whether workers received salaries ($\text{LM-Employee}$); were owners or partners ($\text{LM-Partner}$), or were involved in another form of employment ($\text{LM-Other}$).

The probability of obtaining higher income levels is a key factor in the decision to migrate. To represent this, Karemera et al. (2000) used gross value added (GVA), Mayda (2010) used the average salary of employees, while Ortega and Peri (2009) and Beine and Parsons (2012) considered GDP per capita. The Ecuadorian census does not include salary information, and gross value added data was not available for provinces for the entire study period. Therefore, in order to take into account the concept of different material characteristics in the place of origin and the destination, we used census information on the condition of dwellings, including their structural characteristics, the water supply, the existence of a sewer system and access to electricity. We constructed an index of material conditions ($\text{Relative Index Material WB}$) for provinces up to 1974 by regressing the non-oil GDP per capita on indicators of the

$^{10}$ The variables of physical distance and time were obtained from two sources: Ecuador’s yellow pages (L Dist Y-P) and Google Maps (L Dist Google). We added the Euclidean distance between the capitals in each of the provinces (L Dist Crow). The distances for the Galápagos Islands were calculated by adding the distance to the closest province with an air link (Pichincha and Guayas).
conditions of dwellings, and keeping the estimated coefficients constant.\textsuperscript{11} Again, we lagged such indicator five years before the initial period of the dependent variable, there is a chance that migrants anticipate future changes of income. Models 4 and 5, including destination-time and origin-time fixed effects respectively, remove bias due to omitted variables in the destination (model 4) and origin (model 5) region. In addition, our measurement of material welfare, based on housing characteristics, is much less cyclical than any wages or income variable. We have no information for amenities in our data set. Nevertheless, the variable that we use for capturing material conditions of people is both just a proxy of income and at the same time an indicator of the material living conditions of the population. Given the structure of fixed effects in the empirical models, we account for permanent natural and human made amenities in both origin and destination.

As an additional approximation to control for the level of income and the selectivity of migrants, we used the level of education (No Education, Primary, Secondary and Higher Education). A higher level of education is expected to enable a higher salary to be obtained. Likewise, higher levels of education in the place of origin are expected to be associated with a higher level of migration. Similarly, given that a population’s characteristics determine the propensity to migrate, we considered age cohorts in the regression analyses.

We considered the rate of urbanization (Urbanization rate) as a pull factor for migration. Urbanisation is associated with economic and social development: increasing industrial expansion, higher productivity and salaries, greater probability of finding work and a better quality of life, despite the high level of urban unemployment. Since Alfred Marshall (1890) there is a theoretical framework proving agglomeration economies. The causes of agglomeration economies are addressed by Duranton and Puga (2004), Rosenthal and Strange (2004) and Puga (2010) among many others. Various studies relate empirical findings of a growth augmenting result of various measures of urbanisation (including urban concentration) on countries’ income in the long run (such as Henderson, 2003; Brülhart and Sbergami, 2009).

\textsuperscript{11} In addition to the regression model, we considered other alternatives based on information about dwelling indicators (arithmetic means and principal components). Details of the construction of the index that was finally used and the alternative indices can be found in the Additional Material.
Table 2. Statistical description of variables

| Migration \(k\) | Average | St.Dev. | Min. | Q1 | Q2 | Q3 | Max. | Asymmetry |
|-----------------|---------|---------|------|----|----|----|------|-----------|
| 1982            | 1466    | 4981    | 0    | 50 | 181| 752| 72843| 9.29      |
| 1990            | 1235    | 3050    | 1    | 85 | 225| 847| 34123| 5.43      |
| 2001            | 1306    | 3452    | 1    | 77 | 239| 858| 39511| 6.14      |
| 2010            | 1369    | 3004    | 4    | 107| 335| 973| 23388| 4.06      |

| L. Migrant \(\%\) | Media | Overall | Between | Within | Min | Max | Correlation with L. Migration | Correlation with L. Pobl |
|-------------------|-------|---------|---------|--------|-----|-----|------------------------------|--------------------------|
| P. Pop \(\%\)     | 5.31  | 1.68    | 1.65    | 0.39   | 0   | 9.54| 1                            |                          |
| L. Dist-crow \(\%\) | 12.42 | 1.27    | 1.23    | 0.31   | 8.3 | 15.19| 0.518                       | 1                        |
| L. Dist-Google \(\%\) | 5.92  | 0.69    | 0.69    | 0      | 3.48| 7.39| -0.580                        | -0.112                   |
| L. Time-Google \(\%\) | 5.75  | 0.54    | 0.54    | 0      | 3.71| 6.69| -0.498                       | -0.095                   |
| L. Dist-YP \(\%\)  | 5.94  | 0.69    | 0.69    | 0      | 3.71| 7.39| -0.585                       | -0.122                   |
| L. Time-YP \(\%\)  | 5.56  | 0.61    | 0.61    | 0      | 3.4 | 6.58| -0.512                       | -0.128                   |
| L. Urbanization \(\%\) | 41.9  | 18.66   | 16.62   | 8.25   | 6.85| 20.71| 0.408                       | 0.309                    |
| Agriculture \(\%\) | 45.21 | 19.7    | 14.47   | 8.3    | 9.56| 76.85| -0.383                       | -0.341                   |
| Mines & Quarrying \(\%\) | 0.30  | 0.01    | 0.01    | 0      | 0.03| 12.63| -0.271                       | -0.200                   |
| Manufacturing Ind \(\%\) | 0.02  | 0.05    | 0.03    | 0.01   | 0.14| 5.48 | 0.278                       | 0.393                    |
| Industry Other \(\%\) | 0.06  | 0.01    | 0.01    | 0.01   | 0.05| 0.11 | 0.270                       | 0.322                    |
| Construction \(\%\) | 5.11  | 2.04    | 1.36    | 1.51   | 1.19| 11.46| 0.224                       | 0.140                    |
| Services \(\%\)   | 5.07  | 13.72   | 11.51   | 7.32   | 16.64| 75.03| 0.375                       | 0.220                    |
| LM-Partner \(\%\) | 49.17 | 9.62    | 4.67    | 8.42   | 22.92| 69.65| -0.018                       | 0.195                    |
| LM-Employee \(\%\) | 39.87 | 11.03   | 7.11    | 8.42   | 18.61| 69.13| 0.177                       | -0.031                   |
| LM-Other \(\%\)   | 12.13 | 5.06    | 3.92    | 3.17   | 1.7 | 31.11| -0.332                       | -0.302                   |
| No Education \(\%\) | 16.01 | 9.98    | 5.61    | 8.29   | 2.21| 45.15| -0.136                       | -0.078                   |
| Primary Education \(\%\) | 52.00 | 10.16   | 4.53    | 9.07   | 27.53| 69.94| -0.269                       | -0.245                   |
| Secondary Education \(\%\) | 25.45 | 12.81   | 5.11    | 11.78  | 5.21| 48.92| 0.165                       | 0.142                    |
| Higher Education \(\%\) | 6.45  | 4.96    | 2.77    | 4.1    | 0.35| 22.34| 0.389                       | 0.294                    |
| Pop. 0-4 \(\%\) | 14.09 | 2.53    | 1.53    | 2      | 10.03| 19.79| -0.332                       | -0.428                   |
| Pop. 5-9 \(\%\) | 13.38 | 1.9     | 1.09    | 1.55   | 9.56| 17.3 | -0.337                       | -0.325                   |
| Pop. 10-14 \(\%\) | 12.42 | 1.24    | 0.8     | 0.94   | 8.68| 14.81| -0.276                       | -0.076                   |
| Pop. 15-19 \(\%\) | 10.44 | 0.74    | 0.48    | 0.56   | 8.21| 12.3 | -0.190                       | 0.098                    |
| Pop. 20-24 \(\%\) | 8.87  | 0.97    | 0.8     | 0.55   | 7.13| 11.8 | 0.253                       | 0.002                    |
| Pop. 25-29 \(\%\) | 7.44  | 1.14    | 1       | 0.52   | 5.81| 11.99| 0.218                       | -0.157                   |
| Pop. 30-34 \(\%\) | 6.33  | 0.93    | 0.73    | 0.58   | 4.79| 9.83 | 0.260                       | 0.005                    |
| Pop. 35-39 \(\%\) | 5.54  | 0.75    | 0.44    | 0.62   | 4.29| 8.68 | 0.241                       | 0.134                    |
| Pop. 40-44 \(\%\) | 8.59  | 1.31    | 0.62    | 1.15   | 6.31| 13.4 | 0.259                       | 0.306                    |
| Pop. 50-54 \(\%\) | 5.84  | 1.18    | 0.74    | 0.93   | 3.55| 8.23 | 0.235                       | 0.391                    |
| Pop. 60-64 \(\%\) | 3.93  | 1.13    | 0.87    | 0.71   | 1.78| 6.47 | 0.154                       | 0.361                    |
| Pop. 70+ \(\%\)  | 3.39  | 1.44    | 1.08    | 0.96   | 1.09| 6.59 | 0.136                       | 0.382                    |
| Index Material WB \(\%\) | 6224.9| 2141   | 1709    | 1292   | 2519| 12544| 0.355                       | 0.411                    |

If urbanisation is expected to promote economic growth, it is likely to be associated with higher opportunities and larger migration flows. In addition, as underlined by Rodríguez-Pose and Ketterer (2012), “economic and noneconomic territorial features have been found to be essential elements determining utility differentials, and hence migration incentives of potential movers, across different territories” (p. 536). A significant number of man-made amenities are efficiently produced in urban areas. Thus, cities lead to more opportunities, and consequently spread the “capabilities” a-la-Sen (Sen, 1987), and improve the well-being of individuals. By the same arguments, we would expect a high rate of urbanization in the place of origin to act as a brake on emigration.
Both the dependent variable and the explanatory variables are expressed in logarithms, with the exception of variables that were already expressed as percentages. Consequently, the coefficients are interpreted as elasticities. Table 2 shows the descriptive statistics of the variables considered in the model.

6. Estimation and results

Basic results

In this section, we present the results of estimating the models described in Section 3, using a panel data analysis and considering random effects. A robust estimation of standard errors was made, and we assumed the presence of a potential time correlation between the observations at origin-destination (Models 1 to 3), destination (Model 4) and origin (Model 5) level. The estimated models measure the impact of push and pull variables on migration flows, as well as the characteristics of resident individuals, to control for the selectivity of migrants to a certain extent. Models 4 and 5 report estimates free of bias resulting from different types of multilateral resistance and will be our preferred models. Table 4 presents the results of all models for comparison.

The distance measure considered was the logarithm of the distance between provincial capitals, which was based on the time it takes to travel on the best route by road, according to Google Maps. In all the estimated models, the parameter associated with this variable was negative, as expected. When fixed effects structures were considered the parameter gets bigger, what stressed the importance of space once local specificities are taken into account.

The index that shows relative differences in material welfare was not significant in Model 4. As this specification controls for the destination’s specific circumstances, we could state that lower levels of material welfare in the place of origin do not lead to an increase in migration. On the contrary, in Model 5, controlling for origin specific factors, we observe a positive and significant parameter. The elasticity is 0.16, a result within the range of results for GDP pc in Caragliu et al. (2013) at the international level, much higher than Rupasingha et al. (2015) for wages at the county level in the US, and close to some of the results in Arzhaghi and Rupasingha (2013) for rural urban migration between US counties. Consequently, we can conclude that material welfare acts as a pull factor for migrants in Ecuador.

Variables related to population structure had significant parameters with different signs, which we interpret as a control for the selectivity of emigrants in the place of origin. Model 4 reports lower
migration from regions with younger (between 20 and 24) and older (above 70) residents, but higher migration for regions with higher proportions of residents between 25 and 29. It is more difficult to interpret the results for the destination, as different signs were found for very close age cohorts. However, the signs and, in most cases, the significance were maintained in the different specifications of the gravity model, which indicates that the impact of the age structure on migration is not affected by bias due to multilateral resistance to migration.

The provinces that had high proportions of the population with primary and secondary levels of education were those with the highest levels of emigration. In contrast, provinces with increasing proportions of the population with higher education qualifications did not have differential levels of migration. The education structure in the destination does not appear to be a pull factor or a barrier to migration flows.

In terms of the structure of employment by sector, we found that a highly cyclical sector such as construction is always associated with significant parameters for the place of origin and destination. This confirms empirical evidence in other studies indicating that migration flows are more feasible in periods of expansion and increasing housing availability than in times of recession. For the destination, the manufacturing industry sector was a significant parameter in all the models we considered. The fact that development associated to the growth in manufacturing industry was significant is in favor the Harris-Todaro models of rural-urban transformation associated to development. On the contrary, the estimates did not report a significant impact of the weight of mining industries as a pull factor, what contrasts the role of the oil sector on making the Amazon an attractive destination for migrants.

The structure of the labour market only appears to have a degree of influence at the place of origin. Thus, in Model 4 the “LM-Others” category was associated with a marginally significant, positive parameter. Therefore, structures that could be related to underemployment in the place of origin could be considered push factors.
| L Time Dist Google | Model 1 | Model 2 | Model 3 | Model 4 | Model 5 |
|-------------------|---------|---------|---------|---------|---------|
| -1.387*** | -1.639*** | -1.634*** | -1.639*** |
| (0.0741) | (0.0746) | (0.0889) | (0.0983) |
| Relative Index Material WB | 0.0382 | 0.0651** | 0.0733*** | -0.00614 | 0.163*** |
| (0.0336) | (0.0286) | (0.0282) | (0.0303) | (0.0371) |
| L Urbanization O | 0.890*** | 0.745*** | 0.750*** | 0.721*** |
| (0.0367) | (0.156) | (0.154) | (0.0921) |
| Urbanization rate O | -0.0102*** | -0.0185*** | -0.0185*** | -0.0188*** |
| (0.00310) | (0.00359) | (0.00354) | (0.00295) |
| Pop 10 14 O | 0.0217 | -0.0202 | -0.0223 | -0.00451 |
| (0.0463) | (0.0487) | (0.0481) | (0.0338) |
| Pop 15 19 O | -0.0337 | 0.0171 | 0.0181 | -0.00463 |
| (0.0382) | (0.0409) | (0.0403) | (0.0442) |
| Pop 20 24 O | -0.118*** | -0.162*** | -0.162*** | -0.158*** |
| (0.0423) | (0.0430) | (0.0424) | (0.0501) |
| Pop 25 29 O | 0.198** | 0.163** | 0.160** | 0.167* |
| (0.0838) | (0.0790) | (0.0779) | (0.0924) |
| Pop 30 34 O | -0.0436 | 0.0140 | 0.0156 | 0.0135 |
| (0.0984) | (0.0973) | (0.0959) | (0.104) |
| Pop 35 39 O | -0.0260 | -0.0204 | -0.0240 | -0.0114 |
| (0.0838) | (0.0841) | (0.0830) | (0.111) |
| Pop 49 49 O | 0.0264 | 0.0109 | 0.0125 | 0.0141 |
| (0.0471) | (0.0490) | (0.0484) | (0.0501) |
| Pop 50 59 O | -0.0104 | 0.0302 | 0.0231 | 0.0294 |
| (0.0668) | (0.0688) | (0.0678) | (0.0670) |
| Pop 60 69 O | -0.0589 | 0.0369 | 0.0320 | 0.0494 |
| (0.0933) | (0.105) | (0.104) | (0.0851) |
| Pop 70m O | -0.173*** | -0.249*** | -0.246*** | -0.257*** |
| (0.0795) | (0.0817) | (0.0807) | (0.0771) |
| Primary Education O | 0.0175** | 0.0105 | 0.0105 | 0.0107* |
| (0.00585) | (0.00650) | (0.00641) | (0.00590) |
| Secondary Education O | 0.0277*** | 0.0226* | 0.0226* | 0.0238*** |
| (0.00973) | (0.0119) | (0.0118) | (0.00985) |
| Higher Education O | 0.00380 | -0.0103 | -0.00986 | -0.0138 |
| (0.0133) | (0.0159) | (0.0157) | (0.0150) |
| Mines & Quarrying O | -0.00730 | -0.00642 | -0.00673 | -0.00758 |
| (0.00846) | (0.00991) | (0.00976) | (0.0114) |
| Manufacturing Ind O | 0.00513 | 0.0141* | 0.0136* | 0.0147 |
| (0.00664) | (0.00771) | (0.00761) | (0.00975) |
| Industry Other O | 0.107 | 0.0358 | 0.0272 | 0.0716 |
| (0.0977) | (0.102) | (0.101) | (0.0867) |
| Construction O | 0.0320** | 0.0485*** | 0.0492*** | 0.0388*** |
| (0.0127) | (0.0139) | (0.0137) | (0.0140) |
| Services O | 0.00991 | -0.00660 | -0.00641 | -0.00542 |
| (0.00654) | (0.00747) | (0.00738) | (0.00641) |
| LM-Employee O | -0.00348 | 0.00230 | 0.00213 | 0.00229 |
| (0.00419) | (0.00423) | (0.00416) | (0.00374) |
| LM-Other O | 0.00229 | 0.00860 | 0.00846 | 0.00889* |
| (0.00522) | (0.00569) | (0.00560) | (0.00531) |
| L Population D | 0.579*** | -0.695*** | -0.690*** | 0.669*** |
| (0.0431) | (0.156) | (0.154) | (0.142) |
| Urbanization rate D | -0.00336 | 0.00356 | 0.00345 | 0.00197 |
| (0.00372) | (0.00439) | (0.00432) | (0.00404) |
| Pop 10 14 D | -0.0737 | 0.0325 | 0.0328 | 0.0453 |
| (0.0495) | (0.0500) | (0.0493) | (0.0432) |
| Pop 15 19 D | 0.172*** | 0.107** | 0.105** | 0.0863* |
| (0.0431) | (0.0447) | (0.0441) | (0.0466) |
| Pop 20 24 D | -0.00920 | -0.0513 | -0.0502 | -0.0477 |
| (0.0459) | (0.0427) | (0.0421) | (0.0447) |
| Pop 25 29 D | 0.151* | -0.0597 | -0.0640 | -0.0666 |
| (0.0815) | (0.0784) | (0.0773) | (0.0655) |
| Category                  | Pop 30-34 D | Pop 35-39 D | Pop 49-49 D | Pop 50-59 D | Pop 60-69 D | Pop 70m D | Primary Education D | Secondary Education D | Higher Education D | Mines & Quarrying D | Manufacturing Ind D | Industry Other D | Construction D | Services D | LM-Employee D | LM-Other D | Constant |
|---------------------------|-------------|-------------|-------------|-------------|-------------|-----------|---------------------|---------------------|---------------------|---------------------|---------------------|------------------|-------------|-----------|-------------|-----------|----------|
|                           | -0.0464     | -0.280***   | 0.195***    | -0.229***   | -0.130      | -0.0885   | 0.00606             | -0.0181*            | 0.00403             | -0.0431***         | 0.0174***          | 0.114            | 0.101***    | 0.0298***  | -0.0105**  | 0.00697   |
|                           | (0.0951)    | (0.0862)    | (0.0538)    | (0.0673)    | (0.0864)    | (0.0709)  | (0.00544)           | (0.00936)           | (0.0131)            | (0.0104)           | (0.00511)          | (0.110)         | (0.0146)    | (0.00623)  | (0.00422)  | (0.00493)  |
|                           | 0.176*      | -0.151*     | 0.245***    | -0.254***   | -0.249**    | 0.135*    | 0.00644             | 0.00363             | 0.00136             | -0.0101            | 0.0168**          | 0.0294          | 0.0649***   | 0.00648    | -0.00298   | -0.00125   |
|                           | (0.0952)    | (0.0880)    | (0.0546)    | (0.0688)    | (0.0981)    | (0.0742)  | (0.00605)           | (0.0108)            | (0.0152)            | (0.0120)           | (0.00673)          | (0.111)        | (0.0138)    | (0.00698)  | (0.00432)  | (0.00530)  |
|                           | 0.182*      | -0.155*     | 0.247***    | -0.259***   | -0.254***   | 0.138*    | 0.00651             | 0.00373             | 0.00115             | -0.0106            | 0.0167**          | 0.0293          | 0.0642***   | 0.00688    | -0.00298   | -0.00121   |
|                           | (0.0940)    | (0.0867)    | (0.0539)    | (0.0679)    | (0.0967)    | (0.0730)  | (0.00597)           | (0.0106)            | (0.0151)            | (0.0119)           | (0.00663)          | (0.110)        | 0.0591***   | 0.00689    | (0.00426)  | (0.00524)  |
|                           | 0.206**     | -0.153**    | 0.238***    | -0.235***   | -0.234***   | 0.131**   | 0.00766             | 0.00456             | 0.00211             | -0.0109           | 0.0171***         | 0.0619          | 0.0591***   | 0.00728    | 0.00119    | 0.000368   |
|                           | (0.0852)    | (0.0773)    | (0.0432)    | (0.0665)    | (0.0902)    | (0.0553)  | (0.00703)           | (0.0104)            | (0.0140)            | (0.0115)           | (0.00509)          | (0.124)        | (0.0110)    | (0.00620)  | (0.00319)  | (0.00425)  |
| Fixed effects             | Time        | Time, origin and destination | Time and origin - destination | Destination -time and origin | Origin-time and destination |
| Observ / Nº groups        | 1,600       | 420         | 1,600       | 420         | 1,600       | 420       | 1,620               | 420                 | 1,620               | 420                 | 1,620 | 420|
| Overall R²                | 0.796       | 0.877       | 0.979       | 0.880       | 0.878       |
| Within / Between R²       | 0.404       | 0.796       | 0.473       | 0.891       | 0.473       | 1.000     | 0.555               | 0.890               | 0.531               | 0.891               |

Note: Significance: *:10%; **:5%; ***:1%. Standard errors are given in brackets (robust, and with the possibility of correlation between the various dyadic origin-destination structures). The default categories are Population below 10 years, No Education, Agriculture sector, and LM-Owner Partner.
Finally, we paid special attention to the population variable and to the rate of urbanization. Model 4 showed how provinces increasing in size pushed out larger numbers of the population, which simply confirms a question of scale that is inherent in gravity models. As also expected, greater and increasing rates of urbanization in the place of origin were factors linked to lower levels of migration. The models that control for multilateral resistance to migration to a greater or lesser extent indicated a greater influence of urbanization than observed in Model 1. Model 4, which controls for the heterogeneity of expectations about the destination, showed a negative impact of urbanization in the place of origin that was 84% higher than the parameter in Model 1. In other words, there are fewer reasons to leave provinces with higher rates of urbanization (more and better services, a priori), and there may be different expectations, perhaps because there is better information about destinations.

The destination’s level of population and rate of urbanization are variables with parameters that require deeper reflection. Model 5 reported a significant negative parameter for population, while the urbanisation rate was not significant. This result could be surprising, but is in line with the description in Section 2, which indicated a drop in the rate of population concentration in the most populated provinces, and that “the most populated provinces are not necessarily those that grow most” (INEC, 2012). In fact, 41% of the urban population was concentrated in the three most urbanized provinces (Galápagos, Guayas and Pichincha) in 1982. This proportion continued to be 41% in 2010, due to the notable increase in the rate of urbanization in the other provinces.

Model 1 found a positive, significant parameter for population size. In order to understand such dramatic difference, we follow Baltagi and Griffin (1984) and Pirotte (1999). If we would assume a dynamic relationship, the fixed-effects estimates would capture the short-run impact of the variable, being the pool and random effects estimations a mix of the long (which would be captured by the between estimate) and short estimates. Consequently, Model 1 showed how most populated provinces are recipients of larger population flows. On the contrary, Model 5 reported that most growing provinces are not the ones receiving higher migration flows. We interpreted these results suggesting that the recent process of development in Ecuador is not driving to deepen territorial concentration, particularly due to the growth of medium-sized and small cities in comparison to the large metropolises of Guayaquil and Quito.

Analysis of sensitivity and robustness
We then estimated the specifications of the model using different measures and sub-samples. Next we included the analysis dividing the full sample in two subperiods. Table 4 displays the results of models 4 and 5 for the subperidos 1982-1990 and 2001-2010. We only present the results of distance, the material well being index and the ones for population and urbanisation rate. The
regressions for the 1982-1990 subperiod, where urbanisation was substantially lower, the urbanisation rate was relatively more important to retain population. When we considered variables associated to destination, the process of population concentration was taking place, as population mattered more as pull factor. On the second subperiod, though, urbanisation is less important. These results contrast with the ones obtained for the full sample. In any case, it is very important to take into account that the global model results are not purely an average between subperiods and consequently that the variance between subperiods matters for understanding results of the final model. Consequently the results obtained for the full data set are describing a 40 years story, in a country where urbanisation has boomed and the regional population flows have decreased.

| Table 4. Results by subperiods |
|-------------------------------|
|                               | Model 4  | Model 5  |
|                               | 1982-1990 | 2001-2010 | 1982-2010 | 1982-1990 | 2001-2010 | 1982-2010 |
| L Time Dist Google Relative Index Material WB | -1.694*** | -1.582*** | -1.634*** | -1.706*** | -1.582*** | -1.639*** |
| L Population_O | 1.635*** | 3.339*** | 0.721*** |
| Urbanization rate_O | -0.0898*** | -0.0110 | -0.0188*** |
| L Population_D | 2.920*** | 1.093 | -0.669*** |
| Urbanization rate_D | -0.0710*** | 0.00361 | 0.00197 |

Given that one of the parameters of greatest interest in the study corresponded to the destination’s population and the rate of urbanization at the destination, we then assessed the robustness of the results, excluding the destinations of the provinces of Pichincha and Guayas on the one hand, and the Galápagos on the other. Table 5 shows the estimates of the parameters. It be seen that the signs and significance of the parameters were similar when these provinces were excluded.

Finally we also tested the influence of the Oil production in the Amazon. Even though we controlled for the weight of such sector, we removed from the sample the three provinces where Mining and Quarrying display a stronger role (Napo, which considers Orellana, Sucumbios, Zamora Chinchipe). The results displayed a non significant parameter for population and a positive and significant parameter for the urbanization rate. Even though the Oil producer regions have experienced a huge urbanisation process, they have been attracting decreasing amounts of migrants. Given their particular characteristics, one they are excluded from the sample we observe how the overall growth in urbanisation in the country has acted as brake in the process of population concentration: many rural provinces that in the 70’s had double digit emigration rates have experienced a a joint course of urbanization and retention of population.
Finally, we acknowledge that despite our concerns and the design of our empirical exercise, endogeneity could be affecting our results if there were shocks affecting both migration and population simultaneously, biasing the population coefficient and consequently affecting other parameters. We have performed additional estimates (not reported) removing population as an explanatory variable. These computations show very minor absolute changes in the parameters of the main explanatory variables, and no change is observed in their significance.

### Table 5. Sensitivity of parameters associated with the destination’s population and rate of urbanization

|                          | Model 1 |          |          |          |
|--------------------------|---------|----------|----------|----------|
|                          | Total Sample | Without Pichincha and Guayas | Without the Galápagos | Without the Oil Provinces ¥ |
| L Population D           | 0.579*** | 0.508*** | 0.347*** | 0.537*** |
| Urbanization rate D      | -0.00336 | -0.00271 | -0.000497 | 0.00818** |

|                          | Model 5 |          |          |          |
|--------------------------|---------|----------|----------|----------|
|                          | Total Sample |          |          |          |
| L Population D           | -0.669*** | -0.375** | -0.759*** | -0.019   |
| Urbanization rate D      | 0.00197  | 0.000610 | -0.00568 | 0.00977** |

¥ The provinces specialized in oil are Napo (which considers Orellana), Sucumbios, Zamora Chinchipe, as both have an average of employment in this sector above 3%.

### 7. Conclusions

In this study, we analysed determinants of migration flows in Ecuador between 1984 and 2010 by estimating gravity models. To obtain robust parameters with no bias from multilateral resistance to migration, we estimated a range of specifications using different fixed effects structures.

The main results obtained are in line with the empirical literature on different countries. Thus, migration flows were greater between more populated provinces that were close to each other. The relative index material wellbeing raised as a pull factor, as the estimations showed significant parameters for the destination regions. The education structure in the destination does not appear to be a pull factor or a barrier to migration flows, while the sectoral composition of the economy had a significant role: the Construction sector reported significant parameters both at the origin and at the destination, while Manufacturing Industry had a key role attracting migrants, in line with the Harris Todaro transformation models. Labour markets with structures close to under-employment in the place of origin could be considered push factors.
Finally, we paid special attention to the role of population and urbanization. In the descriptive analysis we found that population flows tend to be towards the most populated provinces, but the concentration rate has dropped over time. Consequently, in recent periods the largest provinces have not been those with the most growth. The estimated models confirmed such trend: territorial concentration has slowed due to the growth of medium-sized and small cities. The sensitivity analysis allowed us to see that the urbanization process in the whole country was acting as deterrent factor for population concentration that is favoured. To be clear, we could consider that a process of territorial balance is occurring in Ecuador, in which growth in provinces, associated to urbanization, is hampering territorial concentration.

In terms of economic policy, the results highlight the importance of understanding jointly the migration and urban phenomenon as a factor that shapes the distribution of a population in space. Consequently, the provision of basic resources (including education and health) should be increased in parallel, or even proportionally more, in small and medium-sized cities. Agglomeration economies could be better exploited if, in practice, increasing levels of urbanization were accompanied by elements that contribute to making better use of the larger size of cities (Castells-Quintana, 2016).

Additional work is required for further understanding of regional migration flows in Ecuador, including the role of international migrations in substituting some internal flows, different behaviours for alternative educational levels, and the inspection of shorter distance flows, such as the ones that take place at the canton level.

8. References
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Appendix 1. Percentage of net migration, population distribution and rate of urbanization. 1982-2010.

| Province          | Percentage of net migration | Provincial weight respect of all national | Urbanization rate |
|-------------------|-----------------------------|------------------------------------------|-------------------|
|                   | 1982 | 1990 | 2001 | 2010 | 1982 | 1990 | 2001 | 2010 | 1982 | 1990 | 2001 | 2010 |
| **The Andes**     |      |      |      |      |      |      |      |      |      |      |      |      |
| Azuay             | -6.9%| -0.2%| 3.9% | 2.7% | 5.5% | 5.3% | 5.0% | 4.9% | 38.3%| 43.2%| 52.1%| 53.4%|
| Bolivar           | -28.6%| -14.6%| -15.1%| -8.7% | 1.8% | 1.6% | 1.4% | 1.3% | 15.6%| 21.1%| 25.5%| 28.2%|
| Cañar             | -9.6%| -3.2%| 1.9% | 0.5% | 2.2% | 2.0% | 1.7% | 1.6% | 16.2%| 29.3%| 36.5%| 42.0%|
| Carchi            | -27.1%| -12.2%| -13.1%| -8.5% | 1.6% | 1.5% | 1.3% | 1.1% | 37.7%| 40.6%| 47.2%| 50.1%|
| Cotopaxi          | -13.0%| -7.9%| -5.1%| -3.4% | 3.5% | 2.9% | 2.9% | 2.8% | 15.4%| 23.7%| 26.8%| 29.6%|
| Chimborazo        | -17.2%| -8.4%| -9.0%| -3.9% | 4.0% | 3.8% | 3.3% | 3.2% | 28.2%| 32.9%| 39.1%| 40.8%|
| Imbabura          | -11.7%| -3.1%| -1.8%| -0.8% | 3.1% | 2.8% | 2.8% | 2.8% | 37.3%| 48.7%| 50.1%| 52.7%|
| Loja              | -25.2%| -11.4%| -9.3%| -4.3% | 4.5% | 4.0% | 3.4% | 3.1% | 33.4%| 39.5%| 45.3%| 55.5%|
| Pichincha *       | 22.3%| 7.2% | 9.7% | 3.8% | 17.2%| 18.3%| 19.8%| 20.4%| 70.4%| 72.9%| 71.8%| 69.0%|
| Tungurahua        | -6.9%| -2.1%| -1.8%| -0.4% | 4.1% | 3.8% | 3.6% | 3.5% | 36.9%| 41.9%| 42.7%| 40.7%|
| **The Coast**     |      |      |      |      |      |      |      |      |      |      |      |      |
| El Oro            | 3.2% | 6.8% | 1.2% | -0.4% | 4.2% | 4.3% | 4.4% | 4.2% | 63.9%| 70.5%| 76.4%| 77.4%|
| Esmeraldas        | -4.4%| -5.9%| -9.2%| -4.9% | 3.1% | 3.2% | 3.2% | 3.7% | 47.6%| 44.0%| 40.7%| 49.6%|
| Guayas **         | 16.1%| 3.9% | 2.7% | 0.8% | 25.4%| 26.3%| 27.4%| 27.4%| 68.7%| 76.3%| 81.8%| 82.2%|
| Los Ríos          | -12.9%| -6.4%| -5.3%| -2.0% | 5.7% | 5.5% | 5.4% | 5.4% | 32.5%| 37.8%| 50.2%| 53.4%|
| Manabí            | -26.0%| -8.5%| -12.5%| -4.2% | 10.8%| 10.8%| 9.8% | 9.5% | 36.7%| 42.0%| 51.9%| 56.4%|
| **Amazon**        |      |      |      |      |      |      |      |      |      |      |      |      |
| Morona Santiago   | 9.9% | 4.6% | -1.3%| 0.5%  | 0.9% | 0.9% | 1.0% | 1.0% | 23.7%| 28.3%| 33.3%| 33.6%|
| Napo              | 46.5%| 15.0%| -1.3%| 0.0%  | 1.4% | 1.1% | 0.7% | 0.7% | 17.4%| 22.9%| 31.4%| 38.1%|
| Orellana          | 18.3%| 9.0% |         |      | 0.7% | 0.7% | 0.7% | 0.7% | 29.8%| 40.3%|         |      |
| Pastaza           | 17.6%| 17.1%| 12.9%| 9.0%  | 0.4% | 0.4% | 0.5% | 0.6% | 32.5%| 36.2%| 43.5%| 44.0%|
| Zamora Chinchipe  | 18.4%| 16.0%| -1.2%| 0.7%  | 0.6% | 0.7% | 0.6% | 0.6% | 22.7%| 24.6%| 35.6%| 39.6%|
|Sucumbios          | 25.4%| 7.7% | -0.7%|      | 0.8% | 1.1% | 1.2% |      | 26.6%| 38.9%| 41.4%|      |
|Galápagos          | 28.1%| 31.2%| 20.8%| 10.9% | 0.1% | 0.1% | 0.2% | 0.2% | 73.4%| 81.9%| 85.4%| 82.5%|

Source: INEC. Note: Data for 2010: * include Pichincha and Santo Domingo; ** Includes Guayas and Santa Elena.
### Appendix 2. Definition and sources of variables

| Variable | Definition | Source |
|----------|------------|--------|
| **Migration** | People who changed residence in the 5 years prior to the Census of Population and Housing (CPV) (INEC, 2014), from the province of origin j to province of destination k | 1982: CEPAL – CELADE; 1990-2010: National Institute of Statistics and Census (INEC) |
| **Population** | Number of people who live in the province j (L_Pop) | 1974-1982: hard copies of the census (CPV); 1990-2010: INEC |
| **Rate of urbanization** | Proportion of individuals who live in areas delimited as urban in each province (Urbanization rate) | 1982: 1982: hard copies of the census (CPV); 1990-2010: INEC |
| **Distance and time – 1** | Kilometres (and time in minutes) from the capital of the province j to the capital of the province k (L Dist Y-P, L Time Y_P) | http://www.guiatefonica.com.ec/Distancia_entre_ciudades_Ecuador |
| **Distance and time – 2** | Kilometres (and time in minutes) from the capital of province j to the capital of province k (L Dist Google, L Time Google) | Google Maps |
| **Branch of activity of the economically active population (EAP)** | Percentage of the economically active population that is employed in the following sectors: Agriculture, hunting, forestry and fishing (Agriculture); Mining and quarrying (Mines & Quarrying); Manufacturing industries (Manufacturing Ind); Other industries (Industry Other); Construction (Construction); Services (Services). | 1974-1982: hard copies of the census (CPV); 1990-2010: CPV – INEC |
| **Level of education** | Percentage of people who have completed no regular education level (No Education) primary education (Primary Education), secondary education (Secondary Education) and higher education, including postgraduate studies (Higher Education). | 1974-1982: hard copies of the census (CPV); 1990-2010: CPV - INEC |
| **Age** | Percentage of people by age groups. (Pop_0_4 … Pop_70 and over) | Census of Population and Housing, National Institute of Statistics and Census (INEC) |
| **Category of employment** | Percentage of people in each employment category: Employee (LM-Employee); Owner or partner (LM-Partner); Other form of employment (LM-Other) | 1974-1982: hard copies of the census (CPV); 1990-2010: CPV - INEC |
| **Characteristics of the dwelling** | Characteristics of the conditions of the dwelling, used to construct the index of material welfare (Index Material WB) | 1974-1982: hard copies of the census (CPV); 1990-2010: CPV - INEC |