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## ABSTRACT

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### **Creativity over Time and Space\***

Creativity is often highly concentrated in time and space, and across different domains. What explains the formation and decay of clusters of creativity? In this paper we match data on thousands of notable individuals born in Europe between the XI<sup>th</sup> and the XIX<sup>th</sup> century with historical data on city institutions and population. Our main variable of interest is the number of famous creatives (scaled to local population) born in a city during a century, but we also look at famous immigrants (based on location of death). We first document several stylized facts: famous births and immigrants are spatially concentrated and clustered across disciplines, creative clusters are persistent but less than population, and spatial mobility has remained stable over the centuries. Next, we show that the emergence of city institutions protecting economic and political freedoms and promoting local autonomy facilitates the attraction and production of creative talent.

**JEL Classification:** R10, O10, J61, J24

**Keywords:** innovation, agglomeration, political institutions, immigration, gravity

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# 1 Introduction

Creativity is often highly concentrated in time and space, and across different domains. In the XV century, Florence was home to an amazing number of groundbreaking innovators in literature, paintings, sculpture and philosophy. At the turn of the XIX century, Vienna hosted pioneers in paintings, medicine, biology, psychology, philosophy, music, who interacted with each other. Baghdad in the IX century, Antwerp in the XVI century, London in the late XVI century and early XVII century, Paris in the early XVIII century, San Francisco and New York in the past few decades, are other examples of clusters of creativity and innovations in several domains (Banks, 1997; Kandel, 2012).

What explains the formation and decay of such clusters of creativity? Are they driven by local economic conditions, by specific features of local institutions, or by mere chance? More generally, besides from these exceptional clusters, how concentrated are creative activities in time and space? Is there co-agglomeration of creative people in different fields? And most important of all, what general lessons can be drawn from the historical analysis of creative clusters, to foster innovation and the production of creative talent? Given the central role of creativity and innovation in human progress and economic development, knowing the answer to these questions is particularly important.

In this paper we analyze data on European creative elites born in the XI-XIX centuries. We exploit information on the dates and location of birth and death of notable individuals in different creative endeavours (arts, humanities, science and business) throughout Europe. The source for these data is *Freebase.com*, a large data base owned by Google and coded by Schich et al. (2014), that stores information from a variety of publicly editable sources, most notably Wikipedia. After integrating these individual data with additional information scraped from the internet, we match them with a historical data set on European cities and local institutions put together by Bairoch et al. (1988) and Bosker et al. (2013). Our unit of observation is thus a city in a particular century between the XI and XIX centuries.

Notable individuals in creative endeavours are a measure of upper tail talent and human capital. They are more likely to capture radical innovations and creativity, compared to general indicators of human capital. Moreover, these data cover periods and domains where patents data are not available. They are thus suitable to study the formation and decay of creative clusters, something that takes place over the long run.

We consider two main variables. First, the number of famous people born in a city (per 1000 inhabitants) during a century. Births of famous creatives are a measure of the local opportunities for innovation offered to highly talented young individuals. As emphasized by Kubler (1962), radical innovations in arts and science reflect a fortuitous match of individual predispositions with

local opportunities for innovation.<sup>1</sup> From this perspective, location of birth is more informative than location of death, because the external environment has a greater impact when individuals are young than when they are old. The local environment can influence individual creativity in many ways. First, familiar role models and local culture play an important role in encouraging young individuals towards creative endeavours (see also Bell et al. 2017). Second, radical innovations are more likely to emerge in a social and cultural environment open to external ideas and where authority and tradition play a lesser role. Third, social learning from peers and older scholars is an important ingredient for successful innovation, through apprenticeships and formal training, but also through interactions and collaborations. Interactions between creatives who sustain and encourage each other while at the same time competing are particularly important in the early stages of artistic or scientific careers. Quoting from Galenson (2009, p.278) : "Location matters to artists primarily early in their careers, because of the need of contacts with other artists".<sup>2</sup>

Our second variable of interest is the number of famous immigrants, i.e. the number of deaths (per 1000 inhabitants) of famous creatives born elsewhere. This variable reflects mobility late in life, and captures the attractiveness of a locality for famous creatives. This in turn can be affected by opportunities for professional enhancement, or by a market or one's services. Given the breadth of our data in terms of time, geography and disciplines, we don't have information on where these notable individuals did their most important work. We doubt that this is an important omission, however, since any invention reflects ideas and experiences accumulated through a life time.<sup>3</sup>

We start by exploring the temporal and spatial patterns of the data, documenting the following stylized facts. First, births of creative people and famous immigrants are *spatially concentrated*, generally more so than population. They are also *clustered across disciplines*, as in the prominent examples of Florence and Vienna cited above, and this is true also after conditioning on local observables. Hence spillover effects associated with local proximity and/or local factors (observed or unobserved) are important for creative activities, and operate across disciplines and not just within each field. This finding is consistent with the discussion and evidence in Jacobs (1969) and

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<sup>1</sup>"When a specific temperament interlocks with a favorable position, the fortunate individual can extract from the situation a wealth of previously unimagined consequences. This achievement may be denied to other persons, as well as to the same person at a different time. Thus every birth can be imagined as set into play on two wheels of fortune, one governing the allotment of its temperament, and the other ruling its entrance into a sequence. ...By this view, the great difference between artists are not so much those of talent as of entrance and position in sequence. ....Predispositions are probably much more numerous than actual vocations allow us to suppose.....Times of opportunities differ more than the degree of talent. " Kubler (1962), p. 7, 8.

<sup>2</sup>Galenson (2009) contains several examples of the importance of collaborations amongst young artists. For instance, referring to the interaction between Picasso and Braques, Picasso is quoted as saying: "Almost every evening, either I went to Braque's studio, or Braque came to mine. Each of us *had* to see what the other had done during the day". Galenson (2009), p. 37.

<sup>3</sup>Quoting again from Kubler (1962), p. 6: "In the long view, biographies ... are only way stations where it is easy to overlook the continuous nature of artistic traditions".

Glaeser et al. (1992).<sup>4</sup>

Second, births and immigration display *persistence*, but smaller than for population. Cities that are at the frontier of creativity in one period retain an advantage that persists for a while but not indefinitely. Estimating a transition matrix, we also find that persistence of creativity is higher at the bottom of the distribution than at the top. Most small and uncreative cities remain in that condition. But at the top of the distribution there is more reshuffling in creative clusters than for population: while most large cities keep growing and remain large, creative clusters exhibit more change over the centuries. The finding that births and famous immigrants are less persistent than population (a proxy for economic activity), despite being spatially more concentrated, is in line with findings on spatial movements of clusters of innovation (Saxenian, 1994; Duranton, 2007; Kerr, 2010).

Third, the overall spatial proximity of births and the distribution of birth-to-death distances *did not change much* over the centuries. This stability is somewhat surprising, in light of the consolidation of states and the improvements in the means of transportation throughout this period. It suggests that the agglomeration of creative activities is not very sensitive to the cost of transportation and communication, but reflects historically stable forces.<sup>5</sup>

We then ask whether creative clusters are influenced by economic prosperity. A priori the answer is uncertain. On the one hand, local wealth creates a demand for the services of artists, through commissions by wealthy patrons. Moreover, flourishing trade and industry can stimulate inventions and facilitate the emergence of a market for works of art, as for Belgian and Dutch paintings in the XV and XVI century (De Marchi and Van Miegroet, 2006). On the other hand, artists and scientists do not mainly chase power and money. Their primary goal is to gain recognition amongst peers. This is not just because of intrinsic motivation and intellectual curiosity, but also because the success of artists and scholars is invariably determined by peer recognition. As emphasized by Galenson (2006), scholars and artists "realize that influence within their discipline will often help them achieve fame and fortune". Kubler (1962, p. 116), expresses a similar view: "The artist requires more than patronage; he also needs association with the work of others ...engaged on the same problems". Moreover, successful innovation often involves more than solving a recognized problem, but also formulating a new question. This requires going beyond satisfying an existing demand or fulfilling a commission. For these reasons, it is not a priori obvious that successful innovators are attracted where the money is.

With these non-experimental data, we cannot definitively establish the direction of causality between creativity and economic prosperity. Nevertheless, we can exploit the temporal pattern

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<sup>4</sup>Jacobs (1969) emphasizes the significance of urban variety and argues that it nurtures cross-fertilization of ideas. Glaeser et al. (1992) find that urban diversity fosters employment growth in industries.

<sup>5</sup>Using more detailed data on famous European musical composers, O'Hagan and Borowiecki (2010) also find that the patterns of their internal migration and emigration have been remarkably stable since the 14th century.

in the data, using population as an indicator of local prosperity in a distributed lag model (using urban population as an index of prosperity is in line with other historical studies - cf. Bosker et al. (2013, p.1418)). We find no correlation between current or lagged population and famous births or immigrants. This suggests that in general local prosperity does not play an important role in the formation or decay of creative clusters. However there is some heterogeneity across disciplines: changes in births of famous non-performing artists and in deaths of famous immigrants in business are preceded by changes in population in the same direction.

Population size is an imperfect measure of economic prosperity. We thus repeat the analysis with historical data on wages of skilled workers, expressed in grams of silver. Here the sample only includes about 30 major European cities, but time is measured in decades rather than centuries, and for most cities the period goes from 1400 to the mid XIX century. There is no evidence that wages started increasing before famous births or the arrival of famous immigrants. This is true for all disciplines. Conclusions are very similar if we use real wages, which are available for a smaller number of city-periods.

These findings may seem surprising to economists, but they are in line with historical anecdotal evidence. Although there are instances where local wealth and artistic florescence go hand in hand, there are also prominent examples to the contrary. For instance, Banks (1997) writes: "Elizabethan London (the apogee of English Renaissance - ed.) suffered "dearness without scarcity" (inflation); this fell most heavily on the aristocracy and the very poor. Then the wool trade collapsed, England entered "the worst economic depression in history" (Wilson, 1965), and Parliament anxiously debated means of averting a *Bellum Rusticum*." Lopez and Miskimin (1962) describe a similar picture for Renaissance Florence, presenting four pieces of evidence for this period: (a) an increase in the fraction of the poor (b) a decline in the city's population (c) the fact that the Medici bank had around half the capital of the Peruzzi bank in 1340 (d) a decrease in wool production (although Cipolla (1964) points out that silk production grew in the same period). Spanish cities in the XVII century are yet another example of negative correlation between artistic excellence and economic prosperity. Quoting from Kubler (1962), p. 114: "XVII century Spain was an epoch of staggering economic difficulties above which painting, poetry, and the theater flowered imperishably". Similarly, Lopez (1953) points out that the famous painter Goya emerged in an impoverished Spain XVIII century, while in earlier centuries rich Genoa was an example of artistic obscurity.

If not local prosperity, what explains the agglomeration of creative individuals in specific localities at different points in time? The rest of the paper explores the effect of local political institutions, and in particular the institutions typical of free cities. We measure city institutions by the dummy variable "Commune" coded by Bosker et al. (2013) and described in the next section. This variable classifies cities based on their political and administrative institutions at about the beginning of each century - this timing is important for our analysis.

As described by Pirenne (2014) and Parker (2004), communal institutions often evolved from within the city, and were guided by the aspiration of the urban middle classes to gain freedom and independence from external influence (primarily in opposition to the Church or an external Lord). In some cases, autonomy and freedom were granted in order to encourage new settlements during periods of intense migration (Bartlett, 1994). The institutions typical of free cities had several implications of great relevance for the agglomeration of creative activities. First, Communes protected basic economic freedoms and promoted trade. As emphasized by Cox (2017), self-governing cities were often in competition to attract merchants and financial capital, as well as talents and human capital. This made free cities a dynamic social environment, in frequent contact with other trading centers and open to external ideas and innovations. Second, Communes guaranteed freedom of movement and from censorship and other personal freedoms; for this reason they often received exiles escaping religious or political persecutions from elsewhere. "Immigrants would settle in towns because they sought liberty, not simply because they wished to trade." (Bartlett, 1993) - see also Burke (2016). Third, Communal institutions created an inclusive social order, that reinforced civic and social capital and emphasized the importance of the common good over particularistic interests. These cultural traits created a fertile ground for innovative activities that would benefit all, such as the pursuit of knowledge and artistic creations.<sup>6</sup> Creative activities directed towards the common good were also incentivized by the Commune. To consolidate their autonomy and territorial expansions and to strengthen identification with the community, independent cities promoted works of art that could become symbols of the city and enhance its prestige - see for instance Paoletti and Radke (2005), Connell (2002) on Florence and Norman (1999) on Siena. All these features of Communes encouraged creative activities and radical innovations, compared to cities that did not have these special institutions.

We first study the correlation between city institutions and the production of creative talent, measured by births of creative individuals in each century (per 1000 inhabitants). Here we do an event study analysis, where the treatment is becoming a Commune. The main result of the event study is illustrated in Figure 1. Becoming a Commune at the beginning of the century (date 0 in Figure 1) is associated with a 5 percentage point increase in the birth of creative individuals (per 1000 inhabitants) during the current century (an increase of about 25% relative to the average number of famous births), with an additional increase in the subsequent century.

Transitions into Commune status are endogenous. To try and reduce the scope for omitted variables, we also estimate by instrumental variable, exploiting the fact that the diffusion of communal institutions occurred in regional waves. Communes emerged in Northern Italy between the XI and

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<sup>6</sup>Quoting from Brucker (2015), p. 30 "Central to the Commune's function was the premise that its corporate components, representing the interests of particular groups, would reconcile their differences within its ambit and under its guidance. Once defined, the common good (*il ben comune*) was expected to take precedence over the interests of any specific group or constituency."



XII centuries, then they spread to Southern France; nearly at the same time as in Italy, Communes also began to appear in the Flanders and in Northern France (Pirenne 1925). Independent cities emerged in Germany in the XIII century, also in association with migration to the East where imperial control was weaker (Parker, 2004). According to historians, these patterns reflect regional factors that influenced the transition into Commune, such as contagion effects in the aspirations of cities, or a vacuum of regional powers (Parker 2004), or in reverse if regional external threats induced transition into despotic rule (as with the Signorie in Italy), or if the consolidation of central states deprived cities of their autonomy. These regional factors also reflected knowledge spillovers in the design of political institutions. As emphasized by Bartlett (1993), towns imitated successful institutional innovations of others, giving rise to families of urban law.<sup>7</sup> To capture these effects, we instrument Commune with the incidence of Commune in the remainder of the region (defined by current NUTS 1 administrative borders), adapting a strategy introduced by Persson and Tabellini (2009) and Acemoglu et al. (2019) in their analysis of democratic transitions in a panel of countries. The identifying assumption is that, conditional on time and city fixed effects and other covariates, the regional waves of institutional transitions influence city level creativity only through the city political institutions. To make this assumption less restrictive, we also control for regional waves of creativity (measured by the spatial lag of births of notable individuals in the region). These IV estimates confirm the results of the event study and are very robust. On average, births of creative people increase by 10 percentage points or more during the century (almost a 50% increase relative to the average), upon a transition into Commune.

Next, we turn to studying whether the emergence (or disappearance) of Communes is correlated with the flow of famous immigrants. Using bilateral data on city of birth and of death by century, we estimate a gravity model of migration (Frankel and Romer, 1999; Bahar and Rapoport, 2018); the relationship between Commune and immigrants is identified through a diff-in-diff methodology. The main finding here is that economic and political freedoms make a city an attractive destination: becoming a Commune is associated with an increase in the inflow of notable individuals, that almost doubles in size.

A possible concern is that our data suffer from large measurement error. It is possible that information on individuals that were closer to political and economic centers is more readily available. Furthermore, these data treat all these notable individuals equally, without weighting them by their achievements and visibility. To address these concerns, we show that our results are robust to using a similar data set that weight individuals by their influence (Yu et al., 2016), and that covers only the upper tail of our famous creatives. Moreover, as discussed in the next section, our

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<sup>7</sup>One example discussed by Bartlett (1993) is the law code of Cuenca-Teruel, that spread in Southern Spain with the Reconquest of previous Muslim territories during the XII and XIII centuries. Other well known examples are the legal codes of Lubeck and Magdeburg, that were widely imitated in the Baltic region and in Northern Germany in the XIII century.

indicators are correlated with measures of technological innovation collected by Meisenzahl and Mokyr (2011) for the UK in the period leading to the Industrial Revolution.

Our paper is related to a large literature. A strong link is with the important work of Mokyr (1990) on the history of technological creativity, and Mokyr et al. (2002) and Mokyr (2016) on the flow of ideas across Europe, and their role in fostering the major scientific innovations that eventually led to the industrial revolution. Mokyr (2002, 2016) mostly focus on the second half of our sample period, and emphasize the importance of interactions within a European community of intellectuals. As suggested by our empirical results, self-governing cities were an important component of the relatively free environment in which these exchanges thrived, and the mobility of creative innovators between cities that competed for their talents was a critical factor in the emergence of this environment. Our paper is also related to Cox (2017), who argues that local autonomy and economic freedoms were key to European growth in the period leading to the industrial revolution, because they set in motion competition between decentralized localities, that in turn favored commerce and economic growth. Finally, our study is also motivated by the literature on upper tail human capital and the industrial revolution (Mokyr, 2009; Meisenzahl and Mokyr, 2011; Squicciarini and Voigtländer, 2014).

The link between democratic institutions and innovation has been studied in the context of economic growth. In particular, Acemoglu and Robinson (2012) discuss of how bad political institutions can block innovation, and Acemoglu (2008) argues that democracies may be better able to benefit from new technologies than oligarchic societies, by letting individuals with comparative advantage in the new technology enter entrepreneurship. Another set of papers illustrate the positive relationship between institutions and urban development (De Long and Shleifer, 1993; Bosker et al., 2013; Guiso et al., 2016; Angelucci et al., 2017).

A related body of work investigates how specific societal attitudes or specific features of the local environment affect innovation, with findings that are very consistent with ours. In particular, Bénabou et al. (2015a) and Bénabou et al. (2015b) show that religiosity is negatively correlated with indicators of innovation (measured at the country and at the individual level). See also Saxenian (1994), Florida (2005), Falck et al. (2011), Acemoglu et al. (2014), Akcigit et al. (2017a) and Akcigit et al. (2017b).

Our paper is also related to a growing literature on innovation. The first relevant body of work, recently surveyed in Carlino and Kerr (2015), analyzes the connections between agglomeration and innovation (Audretsch and Feldman, 1996; Glaeser, 1999; Carlino et al., 2012). In addition, our study is related to the literature on agglomeration advantages, recently reviewed by Combes and Gobillon (2015), and in particular on the microfoundations of such advantages based on knowledge spillovers (Glaeser et al., 1992; Jaffe et al., 1993; Rosenthal and Strange, 2003; Moretti, 2004; Carlino et al., 2007; Arzaghi and Henderson, 2008; Kantor and Whalley, 2014; Guiso et al., 2015;

Serafinelli, 2019). Most of these papers focus on recent periods, however, and they often exploit patent data; in contrast, we study a very long historical period with data on creative elites.

A similar historical perspective to ours is taken by a set of studies using microdata on upper tail human capital, such as Schich et al. (2014), De la Croix and Licandro (2015) and Gergaud et al. (2016).<sup>8</sup> In particular Gergaud et al. (2016) analyze a database of more than one million famous individuals and more than seven million places associated with them throughout human history (3000BCE-2015AD). They document several interesting facts regarding notable people, including a positive correlation between the number of entrepreneurs and artists and subsequent urban growth, which is consistent with our evidence, and a zero or negative correlation between the share of “militaries, politicians and religious people” and urban growth. Relative to their paper, we focus more specifically on documenting spatial patterns of creative individuals, and on the effects of local self-government institutions on the formation of creative clusters. Finally, two related papers study the effects of local institutions on innovation, using historical data on Germany. Donges et al. (2016) show that counties whose institutions are more inclusive as a consequence of the French occupation after 1789 turn out to be more innovative (in terms of patents per capita). Dittmar and Meisenzahl (2016) show that cities enacting in the XVI century legal reforms that established mass public education and increased state capacity began differentially producing and attracting individuals with upper tail human capital.

The outline of the paper is as follows. The next section defines the data and their sources. Section 3 describes a number of stylized facts about the spatial and temporal distribution of creativity. Section 4 describes the results on the determinants of the birth of famous creatives, and in particular on the role of local institutions. In Section 5 we study more in detail the determinants of the migration of famous people between European cities. Section 6 discusses future directions and concludes.

## 2 Data and Variable Construction.

The data used in this paper cover Europe between the XI and the XIX century. Our data set combines information on notable individuals with population and socio-institutional variables at the city level.

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<sup>8</sup>For related work on creative individuals in music and visual arts (and more recent periods) see Hellmanzik (2010), Borowiecki (2013), Borowiecki (2015) and Mitchell (2016). Scherer (2004) studies the lives of several hundreds of famous musicians, focusing on the transition between patronage and free lance in shaping the market for the services of these artists. See also Murray (2003) for a broader and more descriptive historical analysis of the determinants of human accomplishments in arts and sciences.

**Notable Individuals** The records on notable individuals come from *Freebase.com*, as coded by Schich et al. (2014). Freebase is a "large Google-owned knowledge base that is publicly editable and available under a Creative Commons Attribution (CC-BY) license, which allows for both sharing and remixing of the data" (Schich et al. Supplementary Material, p.2). It stores information from a variety of sources, most notably Wikipedia, and contains dates and locations of birth and death, as well as occupations, of notable people.<sup>9</sup> Location information in the records is geocoded, making good quality latitude and longitude data available. Notability of people is "simply defined as the curatorial decision of inclusion" in the (partly crowd-sourced) Freebase (Schich et al. 2014), p. 558). Using these records, we identify 40,980 notable individuals who can be matched by city of birth and/or city of death to the Bairoch et al. (1988) sample (described below). If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample.<sup>10</sup> Of these individuals, we retain 21,906 who became famous thanks to their creative endeavours in the following domains: arts (performing and non-performing), humanities and science, and business. Table 1 provides a count of the individuals active in each domain. The last row reports the total number of creative individuals.<sup>11</sup> Famous creatives in performing arts include: actors, singers, musician, playwrights; in non-performing arts: writers, novelists, journalists, composers, authors, architects; in humanities and science: mathematicians, physicians, philosophers, scientists, physicists, chemists, historians; in business: entrepreneurs, engineers, business-persons, sailors, managers.<sup>12</sup> The vast majority (86.2%) of the individuals studied did all of their work in a single area. Another 12.6% of our famous creatives had achievements in two fields, while 1.1% of individuals achieved fame in three different categories. A few individuals were prominent in all four fields. A further breakdown of polymaths is available in Table A.1.

Using this information, we define the variable  $Births_{ct}$  as the number of famous creatives born in city  $c$  during century  $t$ , per 1000 inhabitants at the beginning of each century. This variable

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<sup>9</sup>See Schich et al. (2014) for more detail on the nature, acquisition and cleaning of the records on notable individuals.

<sup>10</sup>Less densely populated areas (mostly in Russia) have few cities included in Bairoch et al. (1988) and distance to the closest city can be large. We thus impose a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution); if distance exceeds this threshold the observation is not included in our sample.

<sup>11</sup>The last row of Table 1 does not correspond to the sum of the rows above, because some famous creatives were active in more than one domain.

<sup>12</sup>The remaining individuals in the Freebase.com database but not included in our analysis were either unclassifiable, or operated in domains where creativity is less important (i.e. sports; and governance, which includes politicians, lawyers, judges, diplomats, soldiers, priests, social activists). Regarding theologians and jurists, two categories of individuals whose interests lie at the intersection of philosophy and religion, and philosophy and law, respectively, we looked at some prominent examples such as Thomas Aquinas, Martin Luther, John Calvin; Baldo degli Ubaldi, Huig de Groot, Cesaria Beccaria and verified that they are included in humanities and sciences. The number of unclassified notable individuals from Freebase.com is actually 25,580. For about 10,500 of these unclassified entries, we were able to establish the main area of work by scraping the internet in March 2017.

measures the city production of upper-tail creative human capital. As a measure of attraction of upper-tail human capital, we define the variable  $Immigrants_{ct}$  as the number of deaths in city  $c$  of famous creatives born elsewhere during century  $t$ , also per 1000 inhabitants. For individuals who die in a century different from that of birth, we face a problem. Ideally, we would like to attribute these famous people to the century in which their migration decision was taken. Hence, using the century of death risks erroneously posticipating their migration decision, while using the century of birth risks erroneously anticipating their migration decision. One of our goals is to estimate the effect of a change in city institutions (or other observables) on migrations. Using the century of birth (irrespective of the century of death) minimizes the risk of erroneously attributing to institutional changes outcomes that actually took place earlier.

Table 2 reports descriptive information on the number of famous creatives (unscaled by city population) born and immigrated in all the cities and total city population in our sample - the city sample by Bairoch et al. (1988) described below. The Table shows that there is substantial mobility of famous creatives in each century: the number of immigrants is a large fraction of the number of births, even in earlier centuries.

Our dataset is very broad in terms of time, geography and discipline, but this comes at the cost of some limitations. First, as illustrated in Table 2, we have better records of more recent individuals. Second, according to (Yu et al., 2016, p.7), Freebase editors may have an English Bias and a Western Bias, as well as a gender-bias towards males; furthermore the database may be unsuccessful in recording information on works where the participation of creative groups (e.g. orchestras or research teams in firms) outdoes that of creative individuals. In our regression analysis we always include century fixed effects and city fixed effects, which address these concerns. Third, it is possible that information is more readily available for individuals that were born in (or migrated to) cities that at the time were renowned centers of excellence in their discipline. If so, our data would overweight creative clusters and discount cities that only gave birth to (or attracted) a few famous creatives. This kind of non-linearity would not be a problem, however, since our goal is to describe and explain patterns in clusters of creativity, more than explaining the location of a few isolated innovators. The opposite mismeasurement is also possible: young individuals born in the vicinity of an existing creative cluster are likely to move there in their formative years, and yet we may classify them as not born in the creative cluster because their birthplace is a city nearby (see footnote 21 below on Florence for some prominent examples). A more serious concern would be over-recording of famous creatives who were born or died in important political and economic centers, because this would create spurious positive correlation with some of our variables of interest. This may be an issue for state capitals, but it is less likely to be a problem for the cities that became commune, given that in our analysis we control for city population.

A final limitation is that these data treat all notable individuals equally, without weighting

them by their achievements and visibility. Below we also discuss the robustness of our results to a similar data set that weights individuals by citations. Specifically, Yu et al. (2016) collect records of individuals present in more than 25 languages in Wikipedia, which contain dates and locations of birth, as well as occupations. Compared to the dataset constructed by Schich et al (2014), this dataset is significantly smaller and does not include information on the place of death. However, it is manually verified, and it is enriched with the Historical Popularity Index (HPI), a measure that integrates information on the number of languages in which a biography is present in Wikipedia, the time since birth, and the number of page-views between 2008 and 2013. Moreover, the creatives in Yu et al. (2016) belong to the upper tail of the fame distribution in Schich et al (2014), which also addresses concerns related to city-based determinants of visibility for lesser known individuals. Using these records, we identify 1583 notable individuals who (a) can be matched by city of birth to the Bairoch et al. (1988) sample and (b) became famous thanks to their creative endeavours in the arts (performing and non-performing), humanities and science, and business. We define the variable *Births*, *Yu et al.*<sub>ct</sub> as the HPI-weighted number of famous creatives born in city *c* during century *t*, per 1000 inhabitants at the beginning of each century. The correlation between *Births* and *Births*, *Yu et al.* is 0.48.

Finally, a word on the timing of construction of the records of notable individuals in our sample. Freebase editors are our contemporaries, and this has the advantage of generating some distance between the date of the innovation and the construction of their record (or of the weights used by Yu et al. (2016)). This distance arguably allows for a better assessment of breakthrough ideas that may have been too radical (and therefore not accepted) at the time of conception.<sup>13</sup> The creators of this type of innovation would be more likely to be recorded in posteros editions than by their contemporaries. And conversely, individuals generating fashionable ideas that did not stand the test of time would be less likely to be recorded in posteros editions. All that said, we have compared *Births* with records from the *Index Bio-bibliographicus Notorum Hominum* (IBN), which was compiled from around 3000 biographical sources (mainly dictionaries and encyclopedias) with year of publication between 1600 and 1980 - see De la Croix and Licandro (2015). We define the variable *Births*, *IBN*<sub>ct</sub> as the number of famous creatives in IBN born in city *c* during century *t*, per 1000 inhabitants. Unfortunately the geocoding information available to us covers only 59 cities in our estimation sample.<sup>14</sup> We therefore cannot perform our main estimation analysis with these data. However, it is interesting to note that the correlation between our variable *Births* and the variable *Births*, *IBN* is 0.57.

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<sup>13</sup>The artists van Gogh and Gauguin are two example of creatives whose works were acclaimed only after their deaths.

<sup>14</sup>We downloaded the data from the *Journal of Economic Growth* website.

**Correlation with measures of technological innovation** To assess the relevance of possible concerns about measurement error and reporting bias, we investigate whether the variable *Births* is correlated with other historical measures of local innovation. Meisenzahl and Mokyr (2011) collected data on mechanics and engineers born in UK between 1660 and 1830, and on the patents that they created.<sup>15</sup> Many of these individuals were not great inventors, but rather highly skilled and able craftsmen, who adapted new technologies and provided micro innovations. Almost one third of these 747 innovators in Meisenzahl and Mokyr (2011) are also included in our sample of famous creatives - probably the individuals with greater accomplishments. But interestingly, our variable *Births* is also correlated with the birthplace of the remaining mechanics and engineers, included in Meisenzahl and Mokyr (2011) but not classified as notable individuals in our sample.

Specifically, from the Oxford Dictionary of National Biography we obtained the place of birth of the innovators in Meisenzahl and Mokyr (2011) *not* in our sample, and matched it with the cities in our data set. Let *Inventors* be the number of mechanics and engineers born in a city during a century (per 1000 inhabitants), and *Patents* be their number of patents (also per 1000 inhabitants), both variables restricted to the individuals in Meisenzahl and Mokyr (2011) that are *not* in our data set. We then regress  $\text{Log}(1+\text{Inventors})$  and  $\text{Log}(1+\text{Patents})$  on  $\text{Log}(1+\text{Births})$  plus other city observables corresponding to all the other city covariates described below (namely, the variables *Large state*, *Bishop*, *Archbishop*, *Capital*, *Plundered*, *Commune*, *Population*, *University*). As shown in Figures A.1 and A.2 in the Appendix, that depict the added-variable plots, our variable *Births* is positively and significantly correlated with both dependent variables. Specifically, a 10% change in *Births* is associated with a 1.8% change in *Inventors* and a 1.6% change in *Patents*. Thus, for the UK between the XVII and XIX centuries, our proxy for production of creative talent predicts indicators of local innovation, and in particular of the technological micro-innovations that contributed to the industrial revolution.

**European Cities** City population is measured at about the beginning of each century. The source is Bairoch et al. (1988). This is a wide-ranging database with information on 2,200 European cities that reached 5,000 residents between 800 and 1,800. Given the scarcity of data on notable individuals in the very early part of the sample, we restrict the analysis to the period between the XI and the XIX centuries, interpolating population for the missing century 1100.<sup>16</sup> Information on socio-economic and institutional variables comes from Bosker et al. (2013), who, for a subset

<sup>15</sup>Meisenzahl and Mokyr (2011, p.42) argue that Britain's industrial precocity owed a great deal to "the technical competence of the British mechanical elite that was able to tweak and implement the great ideas and turn them into economic realities".

<sup>16</sup>The sample includes cities belonging to the following present-day countries: Albania, Austria, Belarus, Belgium, Bosnia and Herzegovina, Bulgaria, Czechia, Croatia, Denmark, Estonia, Finland, France, Germany, Greece, Hungary, Ireland, Italy, Latvia, Lithuania, Luxemburg, Macedonia, Malta, Netherlands, Norway, Poland, Portugal, Rumania, Russia, Serbia, Slovakia, Slovenia, Spain, Sweden, Switzerland, UK, Ukraine.

of the cities in Bairoch et al. (1988), assembled a large database covering an extensive array of institutional characteristics of European cities between the IX and the XIX centuries. The sample covered by Bosker et al. (2013) includes all cities in Bairoch et al. (1988) that reached 10 000 inhabitants between 800 and 1,800. In our analysis on the effect of institutions, our sample is always that of Bosker et al. (2013), except in section 3 where we describe the main features of the data and we rely on the larger sample of cities by Bairoch et al. (1988). Note that the data coded by Bosker et al. (2013) seek to capture the institutions that were in place at the beginning of each century. Thus, institutional changes that took place during century  $t$  would generally show up in century  $t + 1$ . In other words, although undoubtedly measured with much timing error, our data are more likely to erroneously postpone institutional changes rather than viceversa, compared to our outcomes of interest (births and immigrations of famous people) - a conservative feature of the data definition, given our goal of estimating the effects of institutional changes.

Our main institutional variable of interest is whether a city had a form of self-governance that gave it some autonomy and could constrain the dominant role of the church, state or feudal lords. This is captured by the dummy variable *Commune<sub>ct</sub>* coded by Bosker et al. (2013). This variable measures the extent of local participatory government at the beginning of each century. Typically Communes had autonomy in the regulation of commerce, taxation and other administrative activities. Communal institutions also guaranteed economic and personal freedoms and enforced the rule of law, also through the evolution of civil and penal codes. Besides check and balances on executive authority, Communal political institutions also had forms of limited representative democracy (Pirenne 1925).<sup>17</sup>

Forms of local participative government began to develop in the X and XI centuries when Europe was politically fragmented, after the fall of the Carolingian Empire. In the power vacuum that ensued, cities could claim a kind of self-rule that was frequently recognized by the sovereign in return for taxes or loyalty (Jones, 2003). This form of self-government emerged in the XI and XII centuries in Southern Europe and spread elsewhere in the following centuries (Figure A.9 in the Appendix). As emphasized by Clarke (1926), communes were constitutional oligarchies that represented the interests of merchants, bankers and landowners. Representatives of a ruling class initially acted as the link between the town and its overlord, and gradually gained autonomy from

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<sup>17</sup>Bosker et al. (2013) rely on several criteria for their classification. First, they check if historical sources mention the presence of communal institutions such as consuls or town councils. The date is then attributed to the whole century subsequent to the first evidence of such institutions. As a fallback option, they use the building date of a town hall, and if this information is also missing, they use information on the first time city rights were granted (as mentioned in historical encyclopedias), dating the commune from the first century after such rights were first granted. The criterion for exit from Commune status is symmetrical, namely they code if local participatory institutions stopped functioning, because the town council was taken over by a powerful local family (as in hereditary Signorie), or it was dissolved by a central authority or external power (again dating it from the first whole century after the occurrence). In the absence of a specific indication of a stop to the town council or of the inclusion of the city into a hereditary Signoria, it is assumed that local participatory institutions kept functioning until 1800, in line with local historical sources.



external influence and became accountable to the city bourgeoisie. The degree of emancipation of cities varied across Europe and by centuries, depending on the strength of control over his territory by the prince. In the kingdom of Naples and Sicily, control was strong enough that communes were rare or non-existent. By contrast, the cities of Northern and Central Italy took advantage of the conflict between the Empire and the Pope to gain autonomy from both sources of power. Similarly, in Germany the landed aristocracy was fully occupied "in resisting or supporting the Emperor, extending boundaries and colonising new lands in the north and east", and this gave German towns an opportunity to grab autonomy and develop their own city institutions (Clarke 1926).

The status of Commune is not irreversible, and we observe transitions in both directions, though transitions into the status of Commune (Figure A.10 in the Appendix) are much more frequent than transitions out of it (Figure A.11 in the Appendix). The XII century is the period with the highest number of transitions into Commune, and the highest incidence of Commune in our sample is observed at the beginning of the XVI century. During the XV century, Communes in Italy started to grant long-term authority to a strongman who then acquired absolutist powers over the city and its territory. Here too, external circumstances played an important role, through emulation or because specific external threats convinced towns to grant extraordinary powers to a single individual. As noted by Guiso et al. (2016): "In several cases the Signoria retained the fundamental institutions of the commune, including the principle that power originated from the people and was to be exercised in the people's name. In cities such as Florence and Genoa, the Signoria also preserved the political institutions and the personal liberties that had characterized the commune period." In this regard the Signoria was an evolution of the Commune (Prezzolini, 1948; Chittolini, 1999). Nevertheless, Bosker et al. (2013) (and hence our data) code the transitions into Signoria as a loss of status of Commune. In other parts of Europe, such as the Netherlands, instead, local lords favoured towns, "granted rights of jurisdiction and administration freely and protected commerce from troublesome neighbours" (Clarke 1926). In our estimation strategy we exploit this geographic variation in the emergence and stability of Communal institutions to build an instrumental variable for being a Commune.

Notable people could also be attracted by (or be born in) cities that had universities or that were the location of political or religious power, although religion also led to persecutions and hence could also expel rather than attract innovative individuals. To capture these features, we rely on the following variables, also in the data set by Bosker et al. (2013): whether a city has a university (*University*), and three variables indicating a city's status in the political and ecclesiastical hierarchy, namely a dummy variable indicating whether a city is the seat of a bishop (*Bishop*), is the seat of an archbishop (*Archbishop*) and is a state capital (*Capital*). We also make use of variables, constructed by Bosker et al. (2013), indicating the number of times a city was plundered in

the previous century (*Plundered*), whether it is ruled by a large state (*Large state*), and whether it belongs to a non-absolutist state (*Non-absolutist state*). Table 3 reports summary statistics for all these variables including births and immigrants for the 2137 cities in our sample<sup>18</sup>.

To study the time series properties and their correlation with economic prosperity, we also collected data on the average nominal wage of skilled worker expressed in grams of silver per day from a variety of sources, as well as data on real wages (i.e. adjusted for purchasing power).<sup>19</sup> The main source is Allen (2001), who in turn relies also on other studies. In addition, we used the data gathered by Bennassar (1999), Boulton (1996), Cabourdin (1968), Feliu (1991), Gibson and Smout (1995), Rappaport (2002), and Scholliers and Avondts (1977). These data are only available for 28 major European cities for a long enough period, but they are yearly and they cover several centuries. Table A.2 in the appendix lists the city and years included in our sample. To reduce measurement error and minimize missing observations, we express the wage as a 10-year average (called *Wage*), and obtain an unbalanced panel by decades that covers the period 1260-1890. Panel (a) and (b) of Figure A.3 show the time variation in wages for the five prominent cities discussed in the subsection right below. Panel (c) show the time variation in real wages, that we use for sensitivity analysis (they are available for a smaller sample). The data display variation, both over time and across cities, particularly for real wages.

Finally, as discussed below, in some specifications we define region-specific or nation-specific variables. They all refer to current administrative borders, as defined by Eurostat. NUTS 1 refers to macro-regions, NUTS 2 to regions. Other geographic variables are defined in context below, when we introduce them.

**Some prominent examples** We now describe in greater detail a few European cities that became amazing creative clusters in specific periods. Our goal in this subsection is to gain a better sense of our data, and also to show that they are consistent with anecdotal and historical evidence on creative cities. Throughout we focus on birth of famous creatives, since this is the variable of main interest.

*Florence and the early Renaissance.* Florence became a Commune during the XII century, and Bosker et al. (2013) code it as a Commune during the XIII-XV centuries, consistently with the principle of dating Commune on the basis of the institutions in place at the beginning of each century. City autonomy was established in two waves during the XII century, and both coincided

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<sup>18</sup>We have 6226 city-century observations; 675 cities and 3091 city-century observations are from the Bairoch et al. (1988) sample.

<sup>19</sup>Skilled workers are typically building craftsmen, carpenters, or, more in general, masters and journeymen. The consumer price index used to obtain real wages is a "Laspeyres index in which the quantity of each good is specified and then the price level computed by valuing those quantities at the prices prevailing in each time and place"(Allen, 2001, p.420)

with the death of a German Emperor (i.e. a period of power vacuum at the center).<sup>20</sup>

Appendix Table A.3 lists the famous creatives that according to Freebase were born in Florence or its vicinity in the period 1100-1499, namely from the century of transition into Commune until it becomes a Signoria. During this period Florence became the cradle of the Renaissance movement. This is reflected in our data. No famous creative is born until 1239, and then there is an impressive acceleration, with the apogee reached during the XV century, when Florence was the city in our sample with the highest number of famous births relative to population (cf. Figure 2).

Several historians have emphasized the important role of civic capital in stimulating Florentine creativity, and its link with Communal institutions. Quoting from Clarke (1926), chp. II : "Renaissance civilisation is primarily an urban civilisation; its greatest contributions to art and literature come from the towns. Florence was the centre of this great movement; by the beginning of the fourteenth century she had already produced the first Renaissance architect of secular buildings, Arnolfo di Cambio; the first great Renaissance painter, Giotto; the first modern historian, Villani; and the first modern poet, Dante. The city was a unit small enough to develop rapidly patriotism, consciousness of individual responsibility and a spirit of emulation, a development which found artistic expression, in a manner unknown in half-organised, half-populated monarchies." - see also Brucker (2015). Other famous creatives born in Florence (or its vicinity) during this period and listed in Table A.3 include the painters Cimabue and Botticelli, the architect Brunelleschi, the explorer Amerigo Vespucci, the historian Niccolò Machiavelli and many others.<sup>21</sup> Communal institutions in Florence also stimulated innovations by preserving a decentralized and competitive market for artists. Quoting from De Marchi and Van Mieghem (2006), p. 74: "... though in the course of the 15th century the Medici family acquired more and more power, Florence remained a republic; there was no Ducal Court. For this reason and because commissioning bodies were many, with a variety of goals, there was no "single dominating authority" around which taste might coalesce."

#### *Antwerp and Amsterdam, and the emergence of a market for paintings*

Communal institutions spread in several Belgian and Dutch cities a few centuries later than in Italy. Bosker et al. (2013), code Antwerp and Amsterdam as a Commune from 1300 and 1400 until the end of the sample, respectively. Here too, transition into Commune was followed by diffuse and important artistic innovations, particularly in paintings. Antwerp is the most creative

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<sup>20</sup>The first official record of the establishment of a Commune in Florence dates 1138 (not long after the death of the German Emperor Henry V); after temporarily losing its independence due to Frederick Barbarossa, Florence became again a free city towards the end of the XII century, taking advantage of the death of Frederick's successor, Henry VI - cf. Najemy (2008).

<sup>21</sup>Note that some famous creatives that spent the early part (or most) of their lives in Florence are not listed in Table A.3. In particular, the historian Villani was born and died in Florence but for some reason he is not included in Freebase. The architect Arnolfo di Cambio was born not far from Florence, but Florence is not the city in Bairoch (1988) closest to his birthplace - the closest city is Poggi Bonci. Likewise, Leonardo is assigned to Pistoia, the closest city to his birthplace (Vinci), although he spent his formative years (and several productive periods) in Florence. Despite such omissions, Florence emerges in our data as an exceptional creative cluster.

city in our sample in the XVI century; Amsterdam is the most creative in the XVII century (and the second most creative in the XVI century). Appendix Tables A.4 and A.5 list the famous creatives in Freebase born in Antwerp (or its vicinity) between 1200 and 1599 and Amsterdam between 1200 and 1699. There is no famous birth until one century after the transition into Commune, while the XVI and XVII centuries stand out as exceptional, particularly for painters. Thus, the well known and remarkable clustering of painters in these two cities is fully captured by our data.

According to art historians, city institutions played an important role in the agglomeration of innovative artists in these two cities, although the mechanism is different than in Florence. Antwerp and Amsterdam were important trade centers. This facilitated the emergence of a market for paintings - cf. De Marchi and Van Miegroet (2006). First, merchants were an important source of demand for paintings - both local merchants as well as foreigners who travelled to the city to exchange other goods. Second, the market and transport infrastructures created to exchange goods were also used in the market for paintings, enabling a local production in excess of the local demand (i.e. Belgian and Dutch paintings were exported all over Europe). Competition amongst different painters for a large and heterogeneous demand, as well as an open and welcoming culture amongst local guilds, encouraged new entries and innovation in varieties. Interestingly, the use of existing market infrastructures to foster a market for paintings was encouraged by city authorities: "In the case of Antwerp the city authorities played a central role, adapting to paintings and prints an older marketing institutions - the display hall, used for textiles at fair times in Bruges. ... A second distinctive feature in Antwerp was the deployment of cloister-like structures known as *panden*, some of them operating for the selling of paintings on a year-round basis. Strikingly, the city authorities took an active role in promoting these dedicated sales venues." - De Marchi and Van Miegroet (2006), p. 86,87. More generally, "Guild openness, civic encouragement, widespread dealing, plus specialization and division of labor practiced in the many crafts making up the painters' guild, and especially between masters' workshops, as well as a marketing and exporting orientation- all these features marked Antwerp as an environment for the production and sale of paintings the like of which had not been seen before", *ibid*, p. 89.

#### *Paris and Vienna in the XVIII and XIX centuries*

Finally, we turn to two more recent creative clusters, Paris in 1700-1899 and Vienna in the second half of the XIX century. Paris is close to the top of the distribution for *Immigrants* in our sample during the XVIII and XIX centuries (less so for *Births*), and Vienna is in the top 5% of the distribution for *Births* in the XIX century.

This is consistent with direct historical and anecdotal evidence. The poet and art critic Guillaume Apollinaire wrote in 1913: "In the XIX century Paris was the capital of the art.". Some decades later, Vienna emerged as the cultural capital of Europe, hosting the Vienna Circle in philosophy, the Vienna school of music with a new generation of composers, the Vienna School of

Medicine, the Vienna school of economics, and large numbers of great architects, artists and scientists. The famous creatives included in Freebase and born (or who died) in Paris and Vienna during this period sum to several hundreds, too many to be listed in a table. Appendix Tables A.6 and A.7 instead report the names of creatives born in these two cities and included in the database by Yu et al. (2016), that only considers the most influential individuals. What is striking in these two tables is not only the number of easily recognizable historical figures, but also the breadth in terms of disciplines.

Anecdotal evidence suggests that important agglomeration effects are at the heart of the success of these two cities in attracting and giving birth to famous innovators. Many painters such as Cézanne, van Gogh and Pissarro expressed the view that having been in Paris was essential for their artistic achievements: "There is a theory that I heard you profess, that to paint it is absolutely necessary to live in Paris, so as to keep up with ideas" (Paul Gauguin to Camille Pissarro, 1881, quoted by Galenson 2009, p. 282). Vienna is even more remarkable than Paris, in that interactions and spillover effects between artists and scientists were commonplace and particularly fruitful. Eric Kandel, winner of the Nobel prize in medicine and also born in Vienna, describes how Viennese painters such as Klimt and Schiele were deeply influenced by the exchange of insights about unconscious mental processes with members of the Vienna School of Medicine. "One of the characteristic features of Viennese life at the time was the continual, easy interaction of artists, writers, and thinkers with scientists. The interaction with medical and biological scientists, as well as with psychanalists, significantly influenced the portraiture of these (..) artists" (Kendal 2012, p. xv). Similar interactions were taking place between musicians, poets, architects, philosophers, scientists. Other famous creatives born in Vienna (or its vicinity) during this period and listed in Table A.7 include the composer Arnold Schoenberg, the philosopher Ludwig Wittgenstein, the psychologist Melanie Klein, the architect Otto Wagner, the physicist Ludwig Boltzmann and many others.

Transition into Commune occurred much earlier for both cities, and so it cannot explain the rise of creativity in these two cities. Nevertheless, the emergence of Vienna as a creative cluster was shortly preceded by national political reforms. "In 1848 Austria's liberal middle class became energized and forced the country's absolute, almost feudal monarchy, .. to evolve along more democratic lines. The ensuing reforms were based on a view of Austria as a progressive, constitutional monarchy...characterized by a cultural and political partnership between the enlightened middle class and the aristocracy. This partnership was designed to reform the state, to support the secular cultural life of the nation, and to establish a free market economy, all based on the modern belief that reason and science would replace faith and religion." - Kandel (2012), p. 8. Vienna was the main beneficiary of these progressive reforms. After the reforms, the city "attracted talented people, especially Jews, from all over the empire...(and) benefitted from an influx of talented in-

dividuals from different religious, social, cultural, ethnic and educational backgrounds" - Kandel (2012), p. 9. Thus, although city institutions were not involved, the example of Vienna tells a similar story as the rise of other creative clusters in earlier centuries.<sup>22</sup>

### 3 Stylized Facts

In this section we explore the spatial and temporal patterns of the data, documenting several stylized facts. Our goal here is not to test specific hypothesis or establish causality, but to describe the main features of the data. Nevertheless, it is useful to be guided by a simple conceptual framework.

As explained in the Introduction, our hypothesis is that births of famous creatives (per 1000 city inhabitants) are explained by the opportunities for innovation that are available to talented individuals in a particular location at a point in time. Broadly speaking, we can think of three forces that shape these opportunities, and thus explain why a talented individual becomes a famous creative. (i) Variables that influence the *local supply* of talented individuals engaging in creative activities. Examples of variables that operate through the supply side are opportunities for education, a political and social environment that encourages innovation, but also wars and negative shocks (eg. persecutions) in neighboring localities that induce an inflow of talented immigrants who in turn can act as role models or teachers for talented natives. In particular, the production and attraction of creative talent and innovators may be positively affected by institutions that protect freedoms and basic human rights and allow open exchange of ideas, such as the variable Commune. (ii) Variables that influence the *local demand* for the services of creative individuals, for instance coming from an enlightened local lord or a rich bourgeoisie, or trade infrastructures that expand demand beyond the local one, or in reverse the saturation of demand for the services of specific forms of arts or inventions.<sup>23</sup> (iii) *Spillover effects* amongst creative elites. As emphasized in the Introduction and in some of the prominent examples discussed above, being close to other creative individuals increases productivity through learning, exchange of ideas, or role model effects - for instance, see Glaeser (1999) and Bell et al. (2017). But spillover effects could also be negative, for instance through pecuniary externalities, if the demand for creative services is predominantly local and creative individuals in similar domains are good substitutes for each other.

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<sup>22</sup>The case of Paris is more complex, since Paris was already a very creative city during the XVIII century, before the French revolution.

<sup>23</sup>Kubler (1962) emphasizes saturation as an important source of dynamics in creative activities across localities. "...every durable and successful form saturates the region of its origin, making it impossible for newer linked forms to occupy the same positions. Around every successful form, furthermore, there arises a protective system of sorts for its maintenance and perpetuation, so that the opportunities for replacement by new design are further reduced in places where older things fill the same need. A living artist often may encounter harder competition from the work of artists dead for fifty years than from his own contemporaries." Instead, "a region with many unfulfilled needs, having the wealth to satisfy them, will under certain conditions attract innovations" - Kubler (1962), p. 117.

**Spatial agglomeration** How concentrated are famous creatives in space? Figure 2 displays the spatial distribution of *Births* in the XV century, the middle of our sample. Darker tones indicate a larger number of *Births*, while population size is captured by the circle diameter. Famous *Births* are shown to be concentrated in a subset of the cities, not always those with larger populations. Amongst the large cities, Florence, Nuremberg and Siena have the most births of creatives, per 1000 inhabitants. These cities are recognized as the centers of the Renaissance in Italy and in Northern Europe respectively. Figure 3 displays the spatial distribution of *Births* in the XIX century, the end of our sample period. Now many more cities are included in the sample, and the darker tones have shifted to Northern Europe and the UK. The spatial distribution of *Immigrants* displays similar patterns - cf. Figures A.7 and A.8: Florence and Rome have the largest number of famous immigrants per capita in the XV century. *Births* and *Immigrants* are positively correlated: their correlation coefficient is 0.56.<sup>24</sup> Thus, cities that give birth to creative individuals also tend to attract famous immigrants - in this sense, one can speak of clusters of creativity. Finally, Figure A.4 in the Appendix illustrates the distribution of *Births* in the XIX century. Most cities have very few or zero births, and the distribution displays a large dispersion. Histograms of other centuries are similar, except that the fraction of cities with strictly zero famous creatives diminishes over time.<sup>25</sup>

How did spatial concentration evolve over time? Figure 4 plots the coefficient of variation of *Births*, *Immigrants* and population between cities, in each century between 1300 and 1800. A higher coefficient of variation indicates more geographic concentration (a plot of Gini coefficients is very similar). Recall that *Births* and *Immigrants* are expressed per capita. The following facts stand out. First, *Immigrants* are always more spatially concentrated than both population and *Births*, presumably due to sorting. Second, until 1600 *Births* were also more spatially concentrated than population. These features suggest that local factors or spillovers associated with spatial proximity were particularly important for creative activities. Third, while early on the tendency for famous creatives has been towards less concentration (convergence) over time, the spatial concentration and the spatial patterns of famous creatives did not change much between 1500 and 1800, despite the consolidation of states and the improvements in the means of transportation, suggesting that the forces behind agglomeration of creative activities are historically stable. Nevertheless, *Immigrants* were more concentrated in 1700, a century of significant innovations in several domains.

<sup>24</sup>Table A.8 reports the correlation between our key variables: *Births*, *Immigrants*, *Population* and *Commune*.

<sup>25</sup>See Table A.9 and A.10 in the Appendix for descriptive statistics for *Births*, *Immigrants* and population in the XVth and the XIXth century. Figure A.5 and A.6 report the share of cities by century with zero *Births* and *Immigrants* respectively. Overall, in the sample around half of city-century observations have zero *Births* and around 70% have zero *Immigrants*.

**Persistence** Next, we analyze the persistence of clusters of famous creatives. A comparison of Figure 2 and 3 suggests that there is some spatial movement of clusters over time. To further explore the temporal patterns of the data, Table 4 displays Spearman’s rank correlation coefficients for *Births*, *Immigrants* and Population over each consecutive century (each row reports Spearman’s  $\rho$  between the same variable measured in  $t$  and in  $t + 1$ ). The Spearman’s  $\rho$  for *Births* and *Immigrants* increases over time but it remains below 0.5 until the last century. *Births* are generally less correlated over time than *Immigrants*, and both are much less auto-correlated than population - though this may also reflect larger measurement error in famous creatives than for population. Table 4 suggests that cities at the frontier of creativity have an advantage, but generally not strong enough to guarantee dominance in creativity in the next century.<sup>26</sup>

To complement this analysis, we estimate Markov transition matrices for *Births*, *Immigrants* and population. Specifically, for each variable (*Births*, *Immigrants* and population) and each century, we partition cities in five groups: the first group includes cities that in a given century had a value of zero for that variable; the remaining groups correspond to the quartiles of the distribution in any given century, conditional on being positive. Table 5 displays the probability that a city transits from the row group to the column group in the next century, estimated by Maximum Likelihood (assuming that transition probabilities have remained constant over time). Thus, with regard to *Births*, the first row says that a city that had 0 *Births* in century  $t$  has a 0.61 probability of retaining 0 *Births* in  $t + 1$ , it has a 0.13 probability of being in the first quartile of cities with positive *Births* in  $t + 1$ , and so on.

For all variables, the top left and bottom right cells in Table 5 are the largest, indicating that the probability of remaining in the bottom and top groups, conditional on being there, is the highest. For cities at the top of the distribution, there is strong evidence of more persistence in population than in famous creatives: a city that belongs to the fourth quartile in population has a probability of 0.79 of remaining there in the next century, while for *Immigrants* this probability is 0.49, and for *Births* it is 0.43. On the other hand, at the bottom of the distribution persistence is roughly similar for famous creatives and for population: cities that have just a few famous creatives and belong to the first quartile have a probability of remaining there or falling in group 0 of about 0.55 for *Births* and *Immigrants*; the corresponding value for population is 0.5. In other words, emerging as a large city or a creative cluster is an unlikely event for cities that start out at the bottom of the distribution: most small and uncreative cities remain in that condition. But at the top of the distribution there is more reshuffling in creative clusters than for population: while most large cities keep growing and remain large, creative clusters exhibit more change over the centuries. This pattern conforms with anecdotal evidence about the rise and decline of creativity in cities like Baghdad, Florence, Rome

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<sup>26</sup>Conclusions are similar if we distinguish between famous creatives in arts (performing and non-performing) and famous creatives in humanities and science plus business.



and Vienna.

Notwithstanding the spatial movement of clusters and the changes in the coefficients of variation displayed in Figure 4, the overall pattern of spatial proximity of creatives is quite stable over time, despite the consolidation of states and the improvements in the means of transportation throughout these centuries. This can be seen in Figure 5 and 6. The former displays the distribution of the distance between places of birth of every pair of famous creatives born in the same century, for different centuries. The latter displays the distribution of birth-to-death distances. Both distributions remained quite stable in different centuries, despite the changes in the cost of transportation and communication during this period.

**Agglomeration across disciplines** We then explore whether creative clusters tend to be specialized or diverse. We want to know whether spillover effects and local factors (observed or unobserved) operate across or within disciplines. We thus estimate the matrix of pairwise correlation coefficients of famous people by discipline, for both *Births* and *Immigrants*. Disciplines are disaggregated in performing arts, non-performing arts, humanities and science, and business. Table 6 reports the results, in Panel A for *Births* and in Panel B for *Immigrants*. In the first line of each matrix, we condition on century dummy variables, to make variables comparable over time; thus, we first regress famous people in each discipline (*Births* or *Immigrants*) on a full set of century dummy variables, and then estimate the correlation coefficients across disciplines of the estimated residuals. In the second line of each matrix we condition on both century dummy variables and the set of observable city characteristics described above.<sup>27</sup> Thus, the second line of each matrix is a measure of co-agglomeration across disciplines due to either unobserved common local factors or spillover effects, while the first row also includes the effect of observables. In Panel A, the dependent variable is defined as  $\text{Log}(1 + \text{Births in discipline } i)$ . We take the Log of  $1 + \text{Births in discipline } i$  (rather than of *Births in discipline*  $i$ ) to retain in the sample the large number of city observations with 0 births - see also the more extensive discussion in the next section. In panel B, the dependent variable is defined as  $\text{Log}(1 + \text{Immigrants in discipline } i)$  in Panel B.

All correlation coefficients are positive and significant and quantitatively large, indicating that creative elites are clustered across disciplines, as in the prominent examples of Florence and Vienna cited above. Correlation coefficients are very similar with or without conditioning on the full set of observables, implying that common observable shocks are not responsible for the correlations. For instance, based on Panel A of Table 6, if we compare two cities with the same observables, but city A has 10 p.p. more *Births* of non-performing artists (per 1000 inhabitants) than city B, then on average city A also has 5.2 p.p. more (scaled) *Births* in humanities and sciences compared to city B. This suggests that spillover effects and/or unobserved local factors operate across disciplines

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<sup>27</sup>They are: *Large state, Bishop, Archbishop, Capital, Plundered, Population, University, Commune*.

and not just within each field.<sup>28</sup> Note that correlations tend to be somehow stronger for *Immigrants* than for *Births*.

**Does population predict famous creatives?** Given the patterns discussed so far, a natural question is whether creative clusters are influenced by local prosperity. As remarked in the Introduction, the answer is a priori uncertain. On the one hand, wealth and prosperity tend to increase the demand for services of creative individuals. On the other hand, the primary goal of artists and scientists is to seek influence and recognition amongst peers; moreover, proximity to other creatives facilitates innovation more than proximity to power and money. With these non-experimental data, we cannot provide a definitive answer. Nevertheless, we can exploit the temporal variation in the data, using city size as an indicator of local prosperity in a distributed lag model that only exploits within city correlations.

Specifically, we estimate the following specification:

$$Y_{ct} = \alpha_c + \delta_t + \pi_1 \log(\text{Population})_{ct} + \pi_2 \log(\text{Population})_{ct-1} + u_{ct} \quad (1)$$

where  $Y_{ct}$  is either  $\text{Log}(1 + \text{Births})$  or  $\text{Log}(1 + \text{Immigrants})$ , and  $\alpha_c$  and  $\delta_t$  are city and century fixed effects. Standard errors are clustered at the level of NUTS 2 regions. We use NUTS 2 level rather NUTS 1 level because of the higher number of groups (269 versus 69 in the full sample) but all results in the paper are similar if NUTS 1 is used. A finding that  $\pi_1$  and  $\pi_2$  are not significantly different from 0 would imply that the formation of creative clusters cannot be predicted by contemporaneous or past levels of prosperity, as captured by city size (or by other indicators described below), casting doubts on the possibility of a causal effect. Note that, this being a rather short panel in terms of number of periods, we do not include the lagged dependent variable, but results are similar if we include it.

The results are presented in Column 1 of Tables 7 and 8 for *Births* and *Immigrants* respectively. We cannot reject the null hypothesis that creative people (born or immigrated) are uncorrelated with current and previous population. In results untabulated here, we have explored heterogeneity across disciplines. Births of famous non-performing artists and deaths of famous immigrants in business are correlated with lagged population, as one would expect if local prosperity increases the demand for non-performing arts and attracts entrepreneurship. In columns 2-4 of Tables 7 and 8 we replace population with each of the three variables indicating a city's status in the political and ecclesiastical hierarchy, *Bishop*, *Archbishop* and *Capital*. Again, we fail to reject the null hypothesis of no positive correlation, except for being a state capital (current and lagged *Capital*

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<sup>28</sup>Since  $\text{Births}_{ct} < 1$  (for the vast majority of observations) and  $\text{Log}(1 + \text{Births}_{ct}) \approx \text{Births}_{ct}$ , we talk of percentage point changes. This is an abuse of language because some cities have  $\text{Births} > 1$ , even though it is true that most observations are between 0 and 1 (see for instance Figure A.4).

are both positive and significant in the *Immigrants* regression, and the cumulative effect of *Capital*,  $\pi_1 + \pi_2$ , is significant at 10% in the *Births* regression). Lagged *Archbishop* has a negative and statistically significant coefficient, suggesting that perhaps being close to religious power may stifle creativity, though the F test for significance of current plus lagged *Archbishop* is only statistically significant for *Births*.

In column 5 we estimate Equation (1) replacing *Population* with the dummy variable indicating whether a city has a form of self governance (*Commune*). Now the coefficient on the current and lagged values of *Commune* are positive and statistically significant in both Tables 7 and 8, suggesting that the failure to reject the null hypothesis in Columns 1-4 is not due to lack of statistical power. The following section investigates more closely the relation between local institutions and creativity.

Finally, in column 6 we include all these variables together on the RHS. Again, besides *Commune*, only being a state capital has a positive and significant correlation with *Births* and *Immigrants*.

This evidence is only suggestive, because time is measured at 100 year intervals, but it is robust. We obtain very similar results when (a) using *Births* rather than  $\log(1 + Births)$  as dependent variable, (b) entering two lags rather than one, (c) allowing the coefficient of population to vary before/after the middle of our sample period. Also, we have used the interaction between a dummy variable indicating whether the city is an Atlantic port and dummy variables from 1500 (Acemoglu et al., 2005) as an alternative proxy for economic success, and reached similar conclusions. Overall, these results suggest that local prosperity (other than being a state capital) does not play an important role in the formation or decay of creative clusters. As remarked in the Introduction, these negative findings are in line with historical anecdotal evidence for Renaissance Florence, Elizabethan London, and Spanish cities in the XVII and XVIII century.

**Do wages predict famous creatives?** To further investigate the dynamics of the relationship between economic prosperity and famous creatives, we also rely on wages of skilled workers. Wages are a better measure of economic prosperity than population. Moreover, they are available at a higher frequency for a subset of European cities, and this is important to detect which variable moves first.

We begin our exploration by estimating the following specification:

$$\log(1 + Unscaled\ Births)_{ct} = \alpha_c + \delta_t + \sum_{k=0}^5 \beta_k \log(Wage)_{ct-k} + \sum_{k=1}^5 \beta_k \log(1 + Unscaled\ Births)_{ct-k} + u_{ct} \quad (2)$$

where  $t$  now represents a decade. Thus, our lag structure goes back 50 years. Because population is not available at the frequency of 10 years, we measure the dependent variable as the sum of famous

births during the decade, not scaled by population (*Unscaled Births*). The sample is an unbalanced panel of 28 cities over the period 1260-1890. Note that here we can include the lagged dependent, despite estimating with a city fixed effect, because the panel is sufficiently long (on average about 30 periods per city).<sup>29</sup> Results are similar when estimating without the lagged dependent variable. Standard errors are clustered at the city level. We also test for first order serial correlation in the estimated residuals, and we can never reject absence of serial correlation.

The results are presented in Column 1 of Table 9: the F test cannot reject that the sum of estimated coefficients of current and lagged wages is equal to zero. Note that failure to reject is not due to lack of statistical power, because the estimated coefficients on the lagged endogenous variable (lagged births) are always highly significant. This regression suggests that wages do not help predict famous births. To further explore this possibility, we estimate a specification with  $\log(Wage)$  as the dependent variable, and the contemporaneous values and leads and lags of  $\log(1 + Unscaled Births)$  on the RHS (for 5 decades):

$$\log(Wage)_{ct} = \alpha_c + \delta_t + \sum_{k=1}^5 \beta_k \log(Wage)_{ct-k} + \sum_{k=-5}^5 \beta_k \log(1 + Unscaled Births)_{ct-k} + u_{ct} \quad (3)$$

If the estimated coefficients on the leads of births were significantly different from zero, this would suggest that shocks to wages in period  $t$  were correlated with famous births in subsequent periods. As shown in column 2, this is not the case. The F-test of joint significance of the lead values of famous creatives born in city  $c$  is not significant. This again suggests that wages are not a leading indicator of subsequent accelerations in famous births.

In Columns 3-4 we repeat the same set of regressions replacing births with the number of famous immigrants born in each decade (again, not scaled by population).<sup>30</sup> The patterns are very similar to Column 1-2.

Overall, the estimates in Table 9 are consistent with the notion that the formation of creative clusters cannot be predicted by contemporaneous or past levels of wages.<sup>31</sup> Conclusions are very similar if we use real wages (equal to the nominal wages divided by the consumer price index), which are available for a smaller number of city-periods (yielding 614 versus 815 for the specifications in Column 1 and 3, and 546 versus 692 for the specifications in Column 2 and 4). In results untabulated here, we have also explored heterogeneity across disciplines, but here the conclusions are different compared to population: results are very similar if we distinguish among famous creatives in the four domains.

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<sup>29</sup>See Judson and Owen (1999).

<sup>30</sup>Results are similar if immigrants are measured by year of death, rather than birth.

<sup>31</sup>In regressions available upon requests, we have also included a distributed lag of immigrants on the right hand sides of (2). The results of interest described above remain substantially unchanged, but lagged immigrants have a positive and significant effect on subsequent births.

## 4 City institutions and birth of famous creatives

This section explores the determinants of births of creative individuals, addressing a specific question: what features of city institutions are associated with the production of creative talent, as measured by births of notable individuals? In particular, we study the role that transitions into and out of the status of Commune play in the births of famous creatives. As already discussed, we expect that more democratic and participatory forms of self-government, protecting economic and political freedoms, favor a more open, inclusive and innovative social environment, and thus are positively associated with births of creative individuals. We also investigate the effect of other institutional changes, such as religious institutions and universities.

### 4.1 OLS Estimates

The regression equation that forms the basis of our empirical analysis in this subsection is:

$$\text{Log}(1 + \text{Births}_{ct}) = \beta_1 \text{Commune}_{ct} + \beta_2 X_{ct} + \beta_3 \text{Spatial\_Lag\_B}_{ct} + \alpha_c + \delta_t + u_{ct}, \quad (4)$$

where  $X_{ct}$  are city-level covariates,  $\alpha_c$  and  $\delta_t$  are city and century fixed effects. Thus this specification implements a difference-in-difference estimation and allows us to include observations with 0 famous births. The covariates  $X_{ct}$  belong to two groups: those less likely to be affected by the status of Commune (*Large state, Bishop, Archbishop, Capital, Plundered*), and those more likely to be influenced by Commune or correlated with the error term, and hence to possibly be "bad controls" (*Population, University*). Exchange of ideas is crucial for successful innovation, and interactions could take place with neighboring areas, and not just within the city. An isolated city is in a very different situation compared to a city located in the middle of a very creative area. To control for these spatial determinants of creativity, we also show estimates including the spatial lag of  $\text{Log}(1 + \text{Births}_{ct})$ , defined as  $\text{Spatial\_Lag\_B}_{ct} = \sum_{d \neq c} \varpi_d \text{Log}(1 + \text{Births}_{dt})$ , where the weights  $\varpi_d$  are the inverse of distance between cities  $d$  and  $c$  within the same NUTS 1 region, and 0 outside the NUTS1 region. This spatial lag thus measures the "creative potential" in the macro region, namely neighboring creativity. It captures possible direct effects of being close to other creative cities, as well as possibly omitted variables correlated with creativity in the vicinity of each city. As discussed below, our instrumental variable approach exploits regional waves of institutional change, and this spatial lag is also important to make the IV exclusion restriction more credible.

Table 10 shows OLS estimates of equation (4). Standard errors are clustered at the region (current NUTS 2) level. In Column 1 we estimate a parsimonious version of the baseline specification where we include period dummies and city fixed effects only. In Column 2, we add the first set

of controls. In Column 3 we add the remaining covariates in  $X_{ct}$ , and in Column 4 we add the *Spatial Lag* of the dependent variable. All columns report a positive and significant coefficient on *Commune*. According to the estimate in column 4, becoming a Commune is associated with a 6 percentage point increase in *Births*, or an increase of 0.7 unscaled births (a 26% increase relative to the average).<sup>32</sup> Regarding the other city-level variables, we find a positive and significant coefficient for (possibly endogenous) *University*, and for *Capital*.<sup>33</sup>

Note that the inclusion of  $\text{Log}(\text{population})$  and of *University* in column 3 does not affect the estimated coefficient of *Commune*, suggesting that the effect of local political institutions is unlikely to go through these two channels. On the other hand, the inclusion of the spatial lag reduces the estimated coefficient of *Commune* from 0.70 to 0.58 in column 4, and the estimated coefficient of the spatial lag of  $\text{Log}(1 + \text{Births}_{ct})$  is large and highly significant, suggesting the possible presence of spatially correlated unobserved determinants of creativity, or positive spillover effects from being close to other creative cities.

Finally, note that transition into Commune status of city  $c$  may indirectly affect creative births also in non-treated cities different from  $c$ . As emphasized by anecdotal evidence and as shown in section 5, when a city becomes a Commune it attracts creative immigrants. Through social learning, this may reduce births of famous creatives in the cities experiencing the outflow. The opposite could also happen, since creative individuals may exert positive spillover effects on neighboring cities. Such general equilibrium effects imply that, even neglecting the identification issues due to unobserved heterogeneity and discussed in subsection 4.3, we cannot interpret the estimated coefficient on *Commune* as an average treatment effect. Note however that some of these general equilibrium effects may be captured by the inclusion of the spatial lag of creative births in column (4) of Table 10. Moreover, adding to the regressors also the spatial lags of famous immigrants into neighboring cities and/or the spatial lag of population does not materially change the numerical value or the statistical significance of the estimated coefficient of *Commune* (results available upon request).

## 4.2 Event Study

A concern for the estimation of the relationship between *Commune* and *Births* arises from the possibility of differential pre-trends. Moreover there may be some interesting post-transition dynamics which are not captured by the estimation procedure in the previous OLS Table. We therefore turn to evaluating the local impact of transition into the status of Commune using an "event-study"

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<sup>32</sup>Since  $\text{Births}_{ct} < 1$  (for the vast majority of observations) and  $\text{Log}(1 + \text{Births}_{ct}) \approx \text{Births}_{ct}$ , the coefficient on *Commune* can be interpreted as a percentage point increase after a transition into commune status. Multiplying the coefficient of *Commune* (0.07) by mean population in the sample (in thousand; 11.8) yields the change in unscaled births.

<sup>33</sup>Religious institutions have a negative estimated coefficient, statistically significant only in some specifications.

research design as in Kline (2011) and Autor (2003), whose exposition we follow here. Such a design allows us to test for the presence of differential pre-trends and recover any dynamics of the Commune effect. We compare changes in *Births* of treated cities (i.e. localities that experience the transition into Commune status) both to cities that have not yet been treated and cities that will never be treated during our sample period.

Specifically, the regression equation is:

$$\log(1 + Births_{ct}) = \sum_{\tau=-2}^T \beta_{\tau} D_{ct}^{\tau} + \alpha_c + \delta_t + u_{ct}, \quad (5)$$

where  $\alpha_c$  and  $\delta_t$  are city and century fixed effects, and  $D_{ct}^{\tau}$  are a sequence of "event-time" dummies that equal one when the transition to Commune is  $\tau$  years away in city  $c$  and  $T$  is the end of the sample period (expressed in event time). Therefore  $\tau = 0$  is the year of transition to Commune and the  $\beta_{\tau}$  coefficients characterize the time path of creativity relative to the date of transition for "treated" cities, conditional on the unobserved variance components  $\alpha_c, \delta_t, u_{ct}$ . We impose the restriction that  $\beta_{\tau} = \bar{\beta}$  if  $\tau \geq 2$  - i.e. the dynamics fades out after two centuries. Moreover, we normalize  $\beta_{-1}$  to zero, so that all post-event coefficients can be interpreted as treatment effects. We estimate (5) by OLS.<sup>34</sup>

Figure 1 displays the estimates. There is no evidence of significant pre-existing trends in *Births* of famous people, before the transition into Commune. Becoming a Commune at the beginning of the century (date 0 in Figure 1) is associated with a 5 percentage point increase in the birth of creative individuals (per 1000 inhabitants) during the current century, with an additional increase in the subsequent century. Results are very similar when adding the same controls as in Table 10, including the spatial lag.

### 4.3 2SLS Estimates

An important limitation of the estimation framework in Figure 1 is the possibility that both local institutions and creativity may be influenced by time-varying omitted factors. For instance, the emergence of a vibrant and successful class of merchants and financiers could induce political transitions into Commune, and also exert a direct effect on the demand for the services of innovative artists. To tackle this challenge, we adapt a strategy introduced by Persson and Tabellini (2009) and Acemoglu et al. (2019) in their analysis of democratic transitions in a panel of countries. Namely, we instrument Commune with the proportion of other cities with Commune status in the region (defined by the current NUTS 1 administrative borders) and in the same century, leaving out

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<sup>34</sup>Among the treated cities, in estimating equation (5) below we remove those for which the status of Commune does not persist until the end of the sample period.

the own-city observation - we call our instrument *Regional Commune*. This instrumental variable exploits the fact that transitions into or out of Commune occur in regional waves. As argued above, this spatial correlation of changes in city institutions reflects learning and spillover effects across neighboring cities, or power vacuum (or consolidation) in higher levels of government that affects several cities in the same region.

The identifying assumption is that, conditioning on all included regressors, regional waves of institutional transitions influence city level creativity only through a city's own political institutions. To make this assumption more credible, the regressors also include regional creativity, measured by the spatial lag of  $\text{Log}(1 + \text{Births})$ . Controlling for the spatial lag of the dependent variable reduces the concern that neighboring cities with strong institutions produce or attract more creatives, which in turn exerts direct spillover effects in the region through migration or knowledge diffusion. Nevertheless, the identifying assumption could still be violated if general equilibrium effects were important.

The 2SLS estimates are shown in Table 11, which also reports a summary of the first-stage and the reduced-form results. The sequence of specifications mirrors that in the OLS Table. *Regional Commune* is always highly significant both in the first stage and in the reduced form regressions (F-statistics for the excluded instrument range from 18 to 23). Table A.11 in the Appendix reports the full first stage estimates - note that the religious variables are generally not statistically significant in the first stage regressions.

On average, upon a transition into Commune, births of creative people (per 1000 inhabitants) increase by about 12 percentage points in the more inclusive specification with the spatial lag. This corresponds to about 1.4 more creative births per century, or about a 47% increase relative to average births. The estimated coefficient of *Commune* is about twice as large as in the corresponding OLS regression. The fact that our IV strategy produces larger effects of city institutions on creativity may reflect attenuation bias in the OLS estimates due to measurement error in *Commune*. Another possibility is that the effect of political institutions is heterogeneous across cities. If so, then consistent OLS estimates the average effect of *Commune* on creativity across all cities. On the other hand, 2SLS estimates the average effect for the cities that are marginal in the transition, in the sense that they become communes if and only if there exists a regional wave of institutional change.<sup>35</sup> If the effect of *Commune* on creativity is larger for cities that are marginal in the transition, the 2SLS estimates exceed those of consistent OLS.

Also note that the results are not affected by inclusion of city size on the RHS, which is never statistically significant, although we know from Bosker et al. (2013) that becoming a Commune is also associated with an increase in city size. This reinforces our previous claim that the formation of creative clusters does not seem to operate through local economic prosperity. The coefficient

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<sup>35</sup>See Imbens and Angrist (1994) for a discussion. For a recent example, see Eiseensee and Strömberg (2007).



estimates of the other institutional variables are very similar to the OLS estimates (not shown).

To further reduce the concern that contemporaneous general equilibrium effects or omitted variables correlated with the instrument may violate the identifying assumption, we have also used *Regional Commune* lagged by one century as the instrument; the coefficient of *Commune* remains significant. The estimates are also not materially changed if the spatial lag of famous immigrants or of population are included amongst the regressors. These results are available upon request.

Overall, the evidence is consistent with the idea that becoming a Commune, and enjoying the resulting autonomy and economic and political freedoms, spreads a culture of openness that encourages innovation and creativity in arts, sciences and business.

## 4.4 Sensitivity Analysis

The main empirical result in this Section is that city institutions protecting economic and political freedoms explain the production of creative elites. We now investigate the robustness of the estimates.

**Alternative measures of creativity.** In our data, being famous is equivalent to being included in the database *Freebase.com*. Yu et al (2016) have created a similar database that weights individuals by their influence (see section 2). Table A.12 in the Appendix replicates the 2SLS estimates of Table 11, replacing the dependent variable *Births* with the corresponding weighted variable obtained from Yu et al (2016). The estimated coefficient of *Commune* is positive and significant across all different IV specifications, although the size of the implied estimated effect of a transition into Commune is a bit smaller than with the unweighted data used in Table 11 (expressed in percent of the mean of the dependent variable).<sup>36</sup>

**Specification and Sample Restrictions** In results untabulated here we have explored the sensitivity of the estimates to alternative specifications and sample restrictions. First, we have included the interaction between a dummy variable indicating whether the city is an Atlantic port and dummies from 1500 onwards, as in Acemoglu et al. (2005). Second, we eliminated city-year observations with unusually high values of *Births* (trimming observations above the 99% percentile). Results are largely unchanged.

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<sup>36</sup>The conclusions regarding the role of local institutions are also unchanged when obtaining the OLS estimates using this weighted dependent variable. Regarding the event study, while the visual pattern is similar to that in Figure 1, the individual  $\beta_\tau$  coefficients are not estimated very precisely. We therefore performed a more formal test of the null hypothesis that the transition into *Commune* has no impact on local creativity: we tested the hypothesis about the average of the  $\beta_\tau$  coefficients for  $\tau = 0$  and  $\tau = 1$ . The estimated average increase over the two centuries starting with the year of the opening is 13 p.p. and significant at 10%.

In the previous analysis we exploited all transitions into and out of *Commune*, estimating an average effect. In Table A.13 we estimate the effect of transitions in the two directions separately. Thus, when estimating the effect of entry into *Commune*, we drop the city-century observations following a negative transition (from *Commune* = 1 back to *Commune* = 0). And when studying the effect of exits, we drop the city-year observations prior to a positive transition (from *Commune* = 0 to *Commune* = 1). The OLS estimates remain very similar to those reported in Table 10, with a p-value below 1%, suggesting that the effect of *Commune* on *Births* is symmetric for transitions on both directions. When estimating by 2SLS, the estimated coefficient of *Commune* remains similar to the IV estimates reported in Table 11 (and is larger for transitions into *Commune* than out of it) but the standard errors increase. The estimated coefficient of *Commune* is significant at 10% for the positive transitions, and loses significance for the negative ones.

A natural question is whether our results are due to a particular period in history, or to a specific set of countries. To answer, we have dropped the earliest centuries (XI, XII and XIII altogether) or the most recent one (XIX) and the results are very similar. However when we start dropping the XIV century (in addition to the three earliest centuries) or the XVIII century (in addition to the most recent one) the coefficient of *Commune*, while positive, is no longer significant. We have also included the interaction between *Commune* and a dummy for the period from 1400 onwards (instrumented with the interaction between *Regional Commune* and this dummy). The coefficient of *Commune* remains positive and significant while the interaction is not significant. Similar results were obtained with a dummy for the period from 1500 onwards. These results suggest that the results do not differ much across centuries, but that the central period 1300-1799 is particularly important for the observed correlations.

We have also dropped (individually) countries representing at least 5% of the sample (France, Germany, Italy, Spain, UK) and pairs of countries representing macro-regions (Spain and Portugal, France and Germany). The coefficient of *Commune* remains positive and significant. When dropping the 7 countries belonging to Eastern Europe, the coefficient of *Commune*, while losing significance, is not very different from the one on the full sample (equal to 0.083 with standard error of 0.051). This suggests that our results are not driven by a particular geographic area, but the positive effect of *Commune* on births of famous creatives is present throughout Europe. All these results are available upon request.

**Poisson Estimation** The log-linear specification described above has several advantages. OLS is the best linear unbiased estimator, its consistency properties are transparent and we can easily estimate also by instrumental variables. Moreover, scaling the dependent variable by *Population* reduces concerns about omitting an important regressor, or viceversa including an important "bad control" - Bosker et al. (2013) show that transitions into *Commune* have significant positive effects

on city *Population* during the same century. Nevertheless, a possible problem with the log-linear specification is the large number of zero observations in *Births* (about half of the overall observations have 0 births - see Appendix Figure A.5). To cope with it, here we also estimate by QMLE Poisson, conditional on the same fixed effects described above. Thus, the dependent variable is the number of famous creatives not scaled by population.

Here the concern that *Population* is an endogenous regressor is important, because the dependent variable is unscaled and hence the error term is likely to be correlated with *Population*. To avoid including a "bad control", rather than controlling for  $\text{Log}(\text{Population})$  as a regressor, we include a set of dummy variables that classify cities according to their place in the size distribution or according to their size. These dummy variables are more time invariant than *Population*, and hence they are less likely to suffer from the "bad control" problem that plagues *Population*, and yet their inclusion makes cities of different size comparable. We estimate with two different definitions of the set of dummies. First, we enter separate dummies for belonging to each of the deciles going from the first to the ninth (in the overall sample of observations), plus one dummy for belonging to the set of percentiles from the 90th to the 94th and one for the percentiles 95th to 98th; thus, the default group consists of cities belonging to the 99th percentile. This specification groups cities so that the first 9 groups have roughly the same number of observations. The last decile would include observations that are very heterogeneous in terms of size, because there are few very large cities. To make these cities more comparable within this top decile, we split it in the finer partition described above. In our second and alternative definition, we include a set of dummy variables that classify cities according to the value of  $\text{Log}(\text{Population})$ , irrespective of the frequency in each bin. Specifically we split the range of variation of  $\text{Log}(\text{Population})$  in the entire sample into 10 equally sized intervals, and enter a dummy variable for each interval except the last one (which is the default). Thus, this specification groups cities so that each interval corresponds to cities of roughly similar size and that differ from each other by about the same percentage, irrespective of the frequency distribution.

Table 12, reports these Poisson estimates. Note that, having changed the dependent variable, we redefine the spatial lag accordingly, as the spatial lag of *unscaled* famous births. City and century fixed effects are always included. We also control for all city observables described above (except *Population*) plus the spatial lag of the dependent variable. In column 1 we include the dummy variables based on the frequency distribution of  $\text{Log}(\text{Population})$ , while in column 2 we include the dummy variables based on the values of  $\text{Log}(\text{Population})$ . The estimated coefficient of *Commune* is very similar in both specifications. The estimated coefficient of 0.94 for *Commune* implies that the birth rate of FC in each century is  $\exp(0.94) = 2.6$  times larger in cities that are *Commune*, compared to the others. On average a non-*Commune* city in the sample features about 1.4 births of famous creatives per century, implying that becoming a *Commune* is associated with

an increase of about 2.1 famous births per century. This is much larger than in the OLS estimate of the log-linear specification of Table 11, where we estimated that transitions into Commune are associated with an increase of about 0.7 unscaled births per century.<sup>37</sup> Overall, although these Poisson estimates cannot exploit our instrument for Commune, they confirm the main finding above.

## 5 Migration of famous creatives

In this section we study the determinants of the migration of famous people between European cities, in a gravity model. This section has one goal: to describe how migration is correlated with observable features of European cities, and in particular which institutional features make a city an attractive destination.

### 5.1 Microfoundations

Let  $m_{jit}$  denote the number of immigrants (unscaled by city size) who die in city  $i$  and were born in city  $j$  during century  $t$  (throughout the century refers to the date of birth, as explained above). Also let  $b_{jt}$  denote the number of famous individuals born in city  $j$  during century  $t$ . By definition, we have:

$$m_{jit} = p_{jit}b_{jt} \quad (6)$$

where  $p_{jit}$  is the share of individuals who move from  $j$  to  $i$  in century  $t$ .

The share  $p_{jit}$  is the result of a deliberate decision to migrate. We model it as in the standard Random Utility Model, following Beine et al. (2016) and McFadden (1974). Specifically, let subscript  $k$  denote individuals, and define  $U_{kjit}$  as the utility of individual  $k$  born in  $j$  if he moves to  $i$  in century  $t$ . We assume:

$$U_{kjit} = w_{it} - c_{jit} + \varepsilon_{kjit} \quad (7)$$

where  $w_{it}$  refers to a deterministic component of utility, such as income and other benefits from being in city  $i$ ,  $c_{jit}$  denotes the cost of moving from  $j$  to  $i$  in century  $t$  and  $\varepsilon_{kjit}$  is an individual specific random component of his utility. Note, that, due to our data limitations, we assume that the deterministic component of utility from being in city  $i$ ,  $w_{it}$ , only depends on time and on the destination city, for all individuals irrespective of their origin.

If we assume that  $\varepsilon_{kjit}$  is independently and identically distributed according to an Extreme Value Type-1 distribution, then (6) and (7) imply that the expected number of immigrants from  $j$

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<sup>37</sup>Being *Bishop* has a negative significant coefficient and *Capital* a positive significant coefficient, confirming the inference from OLS regressions that religious institutions are negatively correlated with creative births, while being a state capital has a positive association. The coefficients on the other city-level variables are not significant.

to  $i$  in century  $t$  can be written as - see Beine et al. (2016):

$$E(m_{jit}) = \phi_{jit} y_{it} b_{jt} / \Omega_{jt} \quad (8)$$

where  $\phi_{jit} = \exp(-c_{jit})$  is decreasing in the cost of moving from  $j$  to  $i$ ,  $y_{it} = \exp(w_{it})$  is a measure of attractiveness of location  $i$ , and  $\Omega_{jt} = \sum_l \phi_{jlt} y_{lt}$  is the expected utility from all possible alternatives available to an individual born in  $j$  - including also the decision to stay (corresponding to  $j = l$ ). Thus, the flow of immigrants from  $j$  to  $i$  is higher if city  $i$  is more attractive relative to the average of all other cities (weighted by the cost of moving and including the city of origin), if the city of origin  $j$  has more famous natives (i.e. more potential migrants), and if the cost of moving from  $j$  to  $i$  is lower.

Adding a well behaved error term  $e_{jit}$  to Equation (8), such that  $E(e_{jit}) = 1$ , allows us to estimate the following gravity equation with dyadic data referring to cities of birth and of death for which  $i \neq j$ :

$$m_{jit} = \phi_{jit} y_{it} \frac{b_{jt}}{\Omega_{jt}} e_{jit} = \exp(w_{it} - c_{jit}) \frac{b_{jt}}{\Omega_{jt}} e_{jit} \quad (9)$$

## 5.2 Data and Estimation

Since the dependent variable is a count variable with a very large number of zeros, we estimate equation (9) by Poisson Maximum Likelihood in the sample of cities included in Bosker et al. (2013). Table A.14 in the Appendix summarizes the main features of the dependent variable. There is a very large number of zeros (more than 99% of all observations), and when positive most dyadic observations have only a few immigrants from the same origin city per century. Nevertheless, more than 83% of the cities in our restricted sample received at least one immigrant throughout the period, and several of them, such as Paris and London, received several hundredths overall. The number of destination cities included in the data set ranges from 77 in the XI century to 358 in the XIX century. Note that we discard all the dyadic city-century observations where the origin city-century has 0 births, since the probability of receiving an immigrant from that origin is always zero by construction.

The variables on the right hand side of (9) have the following observable counterparts.

To measure the bilateral cost of moving,  $c_{jit}$ , we use geographic distance (expressed in 100 kilometers), a time varying variable measuring the fraction of each century in which cities  $i$  and  $j$  belonged to the same historical state (Schönholzer and Weese, 2018), and a set of dummy variables that equal 1 if cities  $i$  and  $j$  belong to the same (modern) NUTS1 region, and to two (modern) countries that share the same first official language (Bahar and Rapoport, 2018).

The utility of being in the destination city  $i$ ,  $w_{it}$ , is proxied by population size (that only in this section is measured in 100,000) as a proxy for economic development, and by a set of dummy

variables that capture the most relevant institutional variables in the dataset by Bosker et al. (2013), namely *Commune*, *University*, *Capital*, *Bishop*, *Archbishop*, and *Plundered*. We expect that being a *Commune*, a state *Capital* and having a *University* all make a city a more attractive destination, while having been *Plundered* has the opposite effect. The two ecclesiastic variables have an ambiguous sign: on the one hand the Church was a sponsor of creative endeavours in artistic domains and a source of wealth, but on the other hand it was also a source of discrimination and censorship.

The number of famous births in the origin city  $j$  is measured in logs (to be consistent with the exponential functional form of the Poisson regression (see (9)) and includes all famous people born in city  $j$  during century  $t$ , irrespective of whether or not they died in a different city, since they were all at risk of migrating.

Finally, the so called "multilateral resistance" term  $\Omega_{jt}$  refers to the attractiveness of all the alternative destinations, for an individual born in city  $j$ . This term has no easily observable counterpart. We thus incorporate it as follows. In a first and most restrictive specification, we assume that the only relevant alternative to moving from  $j$  to  $i$  is remaining in the origin city, and thus proxy  $\Omega_{jt}$  with  $w_{jt}$ , namely with the same institutional variables described above but referring to the origin city. Here we also include a full set of century fixed effects, to capture possible symmetric changes in the cost of moving or in the available alternatives. We then relax this assumption by adding also origin and destination fixed effects, to capture possible time invariant omitted variables. Finally, we estimate with a full set of destination and origin-century fixed effects, with which we fully capture the multilateral resistance term  $\Omega_{jt}$  and any other variable that varies by origin and century (as well as time invariant destination variable). Standard errors are always clustered two ways, by origin and by destination, as suggested by Cameron et al. (2011).

The inclusion of destination (and origin) fixed effects implies that we are identifying the parameters of interest with a diff-in-diff methodology. Namely we assume that *changes* in the institutions of interest are randomly assigned to cities, after controlling for the remaining covariates. In particular, we must assume that there is no time-varying unobserved heterogeneity making cities that adopt specific institutions also more likely to attract or send out immigrants (for reasons unrelated to the institutional changes). Note that immigrants are measured by century of birth, to minimize the risk that the migration decision precedes the institutional change. Nevertheless, the identifying assumption remains restrictive.

### 5.3 Results

Table 13 reports the estimates for the three specifications: with only century fixed effects (column 1), with century, destination and origin fixed effects (column 2), and with destination and origin-year fixed effects (column 3). The more credible specification is the one reported in column (3),

but the estimated coefficients of the destination variables remain very stable in columns (2) and (3). Being a *Commune* and a state capital is associated with an inflow of immigrants; the coefficient on *University* is not significant; *Bishop* has a negative coefficient. The estimated coefficients on all the distance measures are highly significant and with the expected sign. Cities that gave rise to more births send out more migrants (the estimated coefficient less than 1 implies that some of famous births do not migrate, as we know from the presence of several natives who die in the city of origin). Finally, population size is not robustly associated with any migration patterns.

The estimated coefficient of 0.520 for *Commune* implies that the arrival rate of immigrants in each century from the same destination city is  $\exp(0.587) = 1.7$  times larger in cities that are *Commune*, compared to the others. On average a non-Commune city receives about 1.6 immigrants per century from all origin cities in this sample, implying that becoming a *Commune* is associated with an increase of about 1.1 famous immigrants per century. Becoming a bishop-city is associated with a drop of about the same size in the immigration rate (though estimated less precisely; p-value is 0.11). Becoming a state capital is associated with an increase in the arrival rate of immigrants of 2.3 times, implying about 2 more immigrants per century.

## 6 Concluding Remarks

It is often argued that open and tolerant political institutions, that protect individual rights and prevent abuse of power by authoritarian leaders, are a prerequisite to sustain innovation-based growth. This argument is strongly supported by the historical evidence of European cities.

As of yet there is no systematic study of the spatial patterns of creativity over a long historical period. After describing the main features of the formation and decay of creative clusters, we study how changes in city institutions affect local creativity. We find that institutions promoting local autonomy and protecting economic and political freedoms encourage the production and attraction of creative talent. The effects are quantitatively large. Becoming a Commune is associated with an increase in the births of famous people of about 40% relative to the average, while the attraction of famous immigrants almost doubles in size upon becoming a Commune. Overall, our estimates strongly suggest that inclusive local institutions and an open environment facilitate the attraction and production of upper-tail human capital in creative occupations.

What are the mechanisms through which becoming a Commune fosters local creativity? We know from Bosker et al. (2013) that transitions into Commune are also associated with subsequent increases in city size. Could this be the mechanism, namely transitions into commune enhance economic prosperity, and this in turn induces an increase in local creativity? Our evidence is inconsistent with this conjecture.

This leaves open the question of what are the mechanisms through which Communal institu-

tions favor the production and accumulation of creative talent. Several answers are suggested by the historical examples briefly discussed in section 2. First, the protection of personal and economic freedoms and an inclusive environment changed the local culture, making it more receptive to innovations and new ideas, enhancing the importance of the common good over particularistic interests, and fostering the appreciation of individual achievements in creative endeavors. Second, the new institutions also changed incentives, through a more meritocratic and less rigid social environment, but also by encouraging works of art and innovations that would enhance the prestige of the city. The Italian Renaissance period exemplifies these two mechanisms. Third, free cities attracted talented and creative individuals who escaped censorship and persecution elsewhere, and this created role models and facilitated social learning, breeding new generations of innovators. Venice, which attracted large numbers of creative immigrants from Greece, Turkey, but also from several European cities, stands out as an example of this mechanism (De Maria, 2010). The inflow of Jews into Vienna from all over the Hasburg empire, after travel restrictions were removed in the mid XIX century, is another example (Weinzierl, 2003). Fourth, the political priority given to the protection of the interests of merchants facilitated the emergence of market infrastructures and exchange networks that could also be exploited for creating a market for works of art. The history of Dutch and Belgian cities, such as Bruges, Antwerp and Amsterdam in the XV, XVI and XVII centuries is an important example. These mechanisms are not mutually exclusive. Discriminating their relative importance and understanding how they operate in different circumstances is an important task for future research, also to assess the external validity of these findings for modern economic development.

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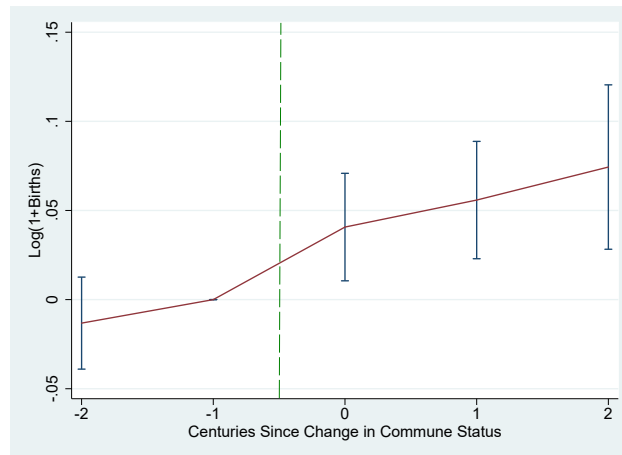
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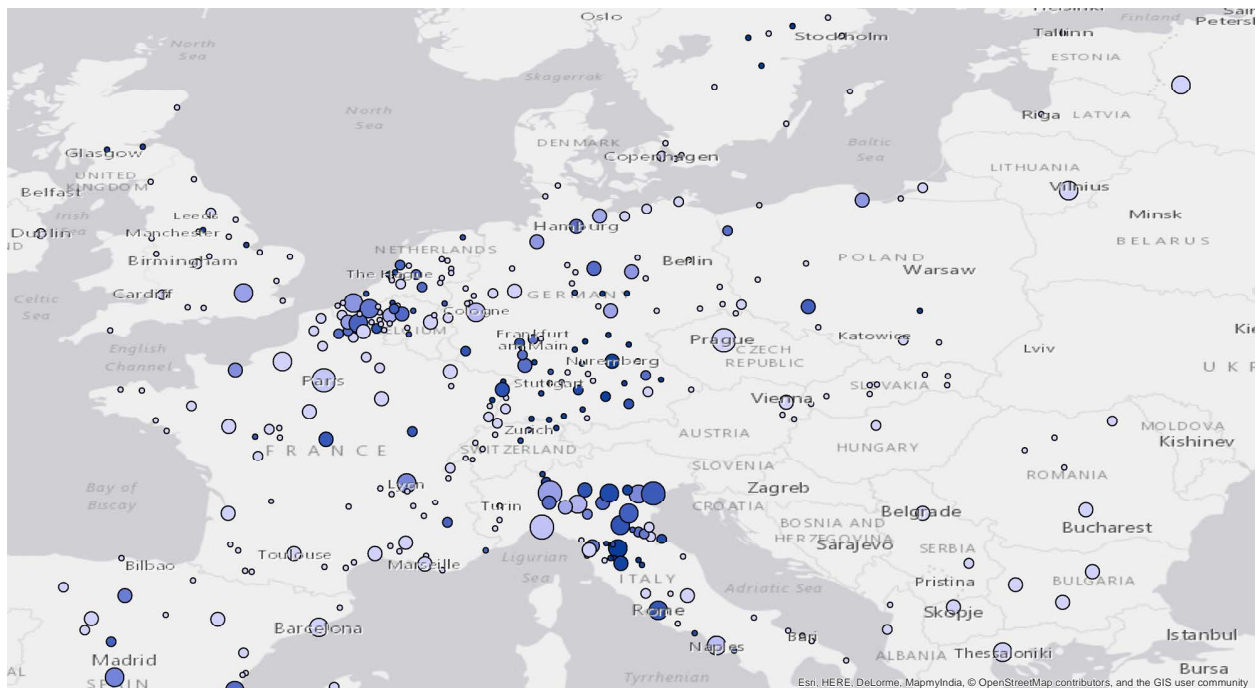
## Figures and Tables

Figure 1: Transitions into Commune and Births of Famous Creatives



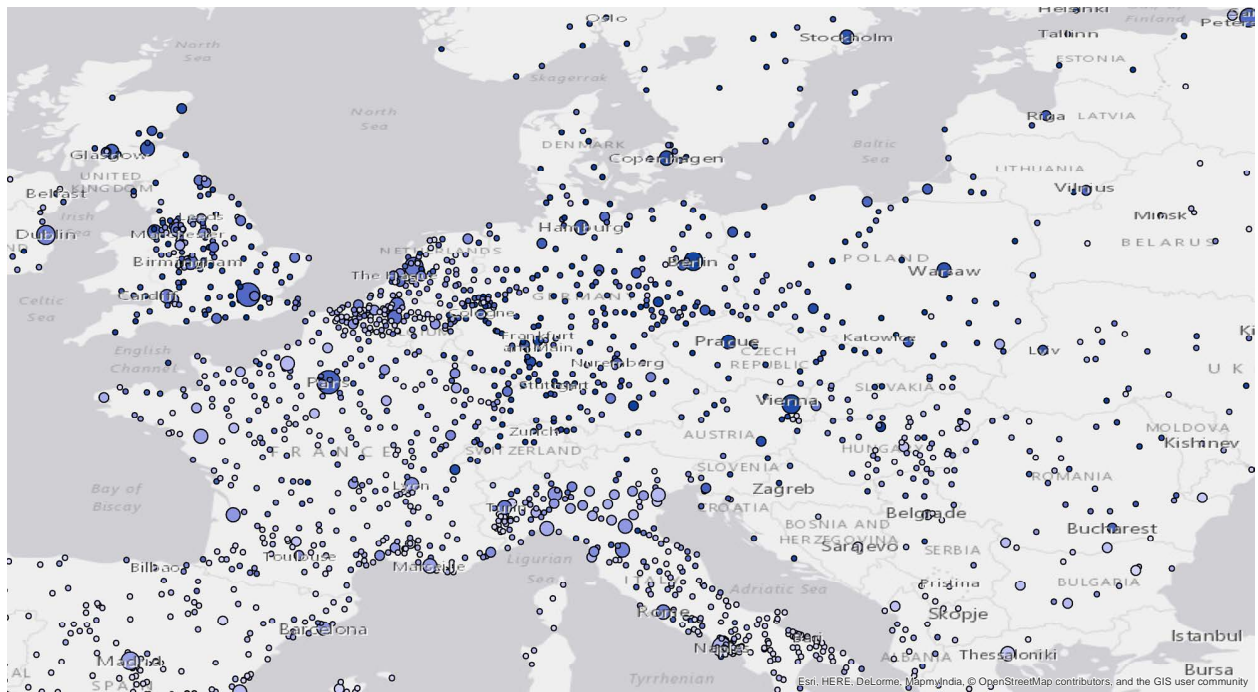
The figure plots point estimates for leading and lagging indicators for the change in commune status. Event time indicator "+2" set to 1 for all periods 2 periods after the event and 0 otherwise. The omitted category is one period prior to the change in commune status. The dependent variable is  $\text{Log}(1 + \text{Births})$ . The variable *Births* is equal to the number of famous creatives born in a city, per 1000 inhabitants. Vertical bars correspond to 95 percent confidence intervals with region-clustered standard errors. We include city FE and century FE. Results are very similar when adding the full set of controls, including the spatial lag.

Figure 2: Spatial Distribution of Births of Famous Creatives, XVth century



The darker the tone, the higher the number of famous creatives born in a city during the century, per 1000 inhabitants. The larger the circle, the larger the population of the city. See Table A.9 for descriptive statistics. The names in the map indicate the location of present-day cities, which may have been small or may have not existed in the XVth century. The map only displays those cities in our sample which are geographically more central.

Figure 3: Spatial distribution of Births of Famous Creatives, XIXth century



The darker the tone, the higher the number of famous creatives born in a city during the century, per 1000 inhabitants. The larger the circle, the larger the population of the city. See Table A.10 for descriptive statistics and Figure 2 for further notes.

Figure 4: Coefficient of Variation of Births, Immigrants, Population, XIV to XIX century

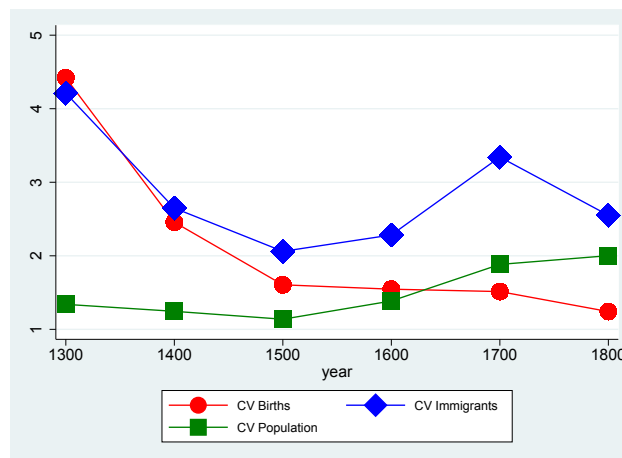


Figure 5: Distribution of Distances between Place of Birth of Any Two Famous Creatives.

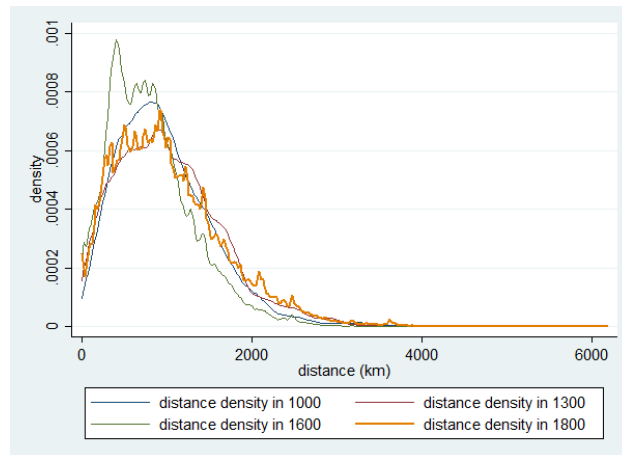


Figure 6: Distribution of Birth-to-death Distances over Time

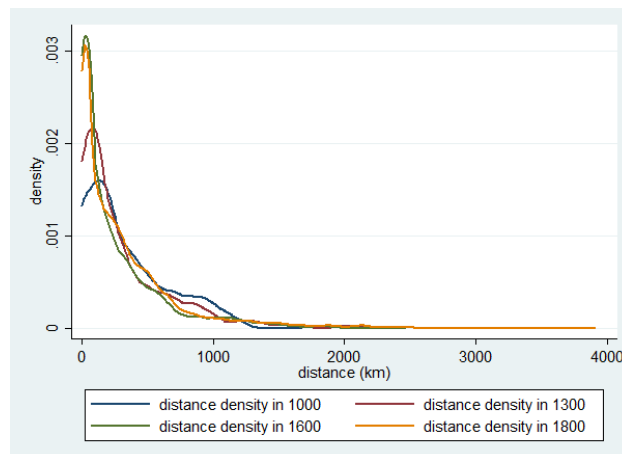




Table 1: Freebase Professional Categories

|                                      | Number of people |
|--------------------------------------|------------------|
| Performing arts                      | 4692             |
| Non performing arts                  | 10002            |
| Humanities and Sciences              | 8402             |
| Business                             | 2083             |
| Total number of creative individuals | 21906            |

The counts refer to the number of people for which a city of birth or death in Europe is observed during our sample period. The total number of creative individuals is lower than the sum across categories because some individuals are present in more than one category.

Table 2: Count of Famous Creatives and Population

| Century | Unscaled Births | Unscaled Immigrants | Population (1000's) |
|---------|-----------------|---------------------|---------------------|
| 1000    | 27              | 20                  | 2004                |
| 1100    | 45              | 33                  | 2089                |
| 1200    | 79              | 45                  | 2993                |
| 1300    | 109             | 79                  | 6092                |
| 1400    | 567             | 346                 | 4867                |
| 1500    | 1243            | 746                 | 6785                |
| 1600    | 1633            | 1068                | 10478               |
| 1700    | 3642            | 2036                | 12595               |
| 1800    | 15027           | 6774                | 25444               |

Table 3: Summary Statistics

| Variable              | Mean      | (Std. Dev.) | Min.    | Max.   | N    |
|-----------------------|-----------|-------------|---------|--------|------|
| Births                | 0.316     | (0.809)     | 0       | 13.25  | 6226 |
| Immigrants            | 0.195     | (1.04)      | 0       | 31     | 6226 |
| (Unscaled) Births     | 3.082     | (16.713)    | 0       | 774    | 6226 |
| (Unscaled) Immigrants | 2.344     | (20.362)    | 0       | 827    | 6226 |
| Commune               | 0.504     | (0.5)       | 0       | 1      | 3091 |
| University            | 0.118     | (0.323)     | 0       | 1      | 3091 |
| Non-Absolutist State  | 0.352     | (0.478)     | 0       | 1      | 3091 |
| Population            | 11780.794 | (25219.232) | 316.228 | 948000 | 6226 |
| Large state           | 0.627     | (0.484)     | 0       | 1      | 3091 |
| Bishop                | 0.382     | (0.486)     | 0       | 1      | 3091 |
| Archbishop            | 0.118     | (0.323)     | 0       | 1      | 3091 |
| Capital               | 0.07      | (0.255)     | 0       | 1      | 3091 |
| Plundered             | 0.022     | (0.146)     | 0       | 1      | 3091 |
| Births, Yu et al.     | 0.925     | (3.607)     | 0       | 99.557 | 6226 |

2137 individual cities; 675 cities are from the Bairoch et al. (1988) sample. Population is interpolated for year 1100

Table 4: Spearman's Rank Correlation Coefficient for Key Variables over Time

| Century           | (1)<br>Births | (2)<br>Immigrants | (3)<br>Population |
|-------------------|---------------|-------------------|-------------------|
| <i>XIV-XV</i>     | 0.164**       | 0.308***          | 0.838***          |
| <i>XV-XVI</i>     | 0.331***      | 0.398***          | 0.870***          |
| <i>XVI-XVII</i>   | 0.424***      | 0.500***          | 0.839***          |
| <i>XVII-XVIII</i> | 0.423***      | 0.462***          | 0.791***          |
| <i>XVIII-XIX</i>  | 0.593***      | 0.516***          | 0.700***          |

Each row reports Spearman's rho for a given variable measured at t and t+1. For instance, the row "XVI-XVII" reports the measure of statistical dependence between the variable in the column measured in XVI century and the same variable measured in the XVII century.

Table 5: Markov Transition Matrices for Key Variables across Cities

|              | Births   |           |           |           |           | Immigrants |           |           |           |           | Population |           |           |           |           |
|--------------|----------|-----------|-----------|-----------|-----------|------------|-----------|-----------|-----------|-----------|------------|-----------|-----------|-----------|-----------|
| <b>t/t+1</b> | <b>0</b> | <b>q1</b> | <b>q2</b> | <b>q3</b> | <b>q4</b> | <b>0</b>   | <b>q1</b> | <b>q2</b> | <b>q3</b> | <b>q4</b> | <b>0</b>   | <b>q1</b> | <b>q2</b> | <b>q3</b> | <b>q4</b> |
| <b>0</b>     | .61      | .13       | .12       | .09       | .05       | .75        | .09       | .07       | .06       | .03       | .61        | .2        | .09       | .07       | .03       |
| <b>q1</b>    | .17      | .42       | .24       | .12       | .04       | .23        | .47       | .19       | .07       | .04       | .25        | .3        | .23       | .19       | .03       |
| <b>q2</b>    | .22      | .22       | .24       | .19       | .12       | .32        | .19       | .15       | .18       | .16       | .1         | .1        | .21       | .44       | .15       |
| <b>q3</b>    | .12      | .08       | .19       | .32       | .29       | .22        | .12       | .16       | .23       | .27       | .05        | .05       | .13       | .35       | .41       |
| <b>q4</b>    | .19      | .03       | .1        | .26       | .43       | .15        | .03       | .1        | .23       | .49       | .01        | .01       | .05       | .14       | .79       |

This Table displays the probability of transition from each row to each column, estimated by Maximum Likelihood. For each variable (Births per 1000 inhabitants, Immigrants per 1000 inhabitants, and Population) and each century, we divide cities in five groups: the first group includes cities that in a given century featured a value of zero (the first row / column). The remaining groups correspond to the quartiles of the distribution in any given century conditional on being positive. The Table displays the probability of moving between each of the five categories from century t to century t+1, estimated by Maximum Likelihood.

Table 6: Coagglomeration

|                         | Arts<br>Non-Performing | Arts<br>Performing | Humanities<br>and Sciences |
|-------------------------|------------------------|--------------------|----------------------------|
| Panel A: Births         |                        |                    |                            |
| Arts, Performing        | 0.4039 ***             |                    |                            |
|                         | 0.3825 ***             |                    |                            |
| Humanities and Sciences | 0.4988 ***             | 0.2757 ***         |                            |
|                         | 0.5229***              | 0.3165 ***         |                            |
| Business                | 0.3327***              | 0.2789 ***         | 0.3386 ***                 |
|                         | 0.3901***              | 0.2472***          | 0.3700***                  |
| Panel B: Immigrants     |                        |                    |                            |
| Arts, Performing        | 0.4700 ***             |                    |                            |
|                         | 0.4471 ***             |                    |                            |
| Humanities and Sciences | 0.5261 ***             | 0.3434 ***         |                            |
|                         | 0.5566***              | 0.2817 ***         |                            |
| Business                | 0.3673 ***             | 0.3944 ***         | 0.2864 ***                 |
|                         | 0.3999 ***             | 0.3381 ***         | 0.3848 ***                 |

Each entry represents a pairwise correlation of residuals from regressions of famous people per capita in each discipline by city of birth (Panel A) or death (Panel B) on period dummies (upper row) and period dummies plus controls (lower rows). The dependent variable is defined as  $\text{Log}(1+\text{Births in discipline } i)$  in Panel A and  $\text{Log}(1+\text{Immigrants in discipline } i)$  in Panel B. \*  $p<0.1$ , \*\*  $p<0.05$ , \*\*\*  $p<0.01$ .

Table 7: Population and Births

|  | (1)               | (2)               | (3)                 | (4)                | (5)                 | (6)                  |
|--|-------------------|-------------------|---------------------|--------------------|---------------------|----------------------|
| Log (Population)                       | -0.016<br>(0.013) |                   |                     |                    |                     | -0.017<br>(0.013)    |
| L.Log (Population)                     | 0.016<br>(0.013)  |                   |                     |                    |                     | 0.015<br>(0.011)     |
| Bishop                                 |                   | 0.004<br>(0.024)  |                     |                    |                     | -0.009<br>(0.027)    |
| L.Bishop                               |                   | -0.037<br>(0.026) |                     |                    |                     | -0.089***<br>(0.031) |
| Archbishop                             |                   |                   | 0.014<br>(0.034)    |                    |                     | 0.017<br>(0.039)     |
| L.Archbishop                           |                   |                   | -0.075**<br>(0.033) |                    |                     | -0.131***<br>(0.037) |
| Capital                                |                   |                   |                     | 0.047**<br>(0.022) |                     | 0.041*<br>(0.022)    |
| L.Capital                              |                   |                   |                     | 0.024<br>(0.023)   |                     | 0.032<br>(0.026)     |
| Commune                                |                   |                   |                     |                    | 0.052***<br>(0.014) | 0.054***<br>(0.017)  |
| L.Commune                              |                   |                   |                     |                    | 0.040***<br>(0.015) | 0.030*<br>(0.017)    |
| $\beta_X + \beta_{L.X} = 0, \text{pv}$ | 0.995             | 0.181             | 0.044               | 0.031              | 0.000               |                      |
| Observations                           | 3,434             | 3,045             | 3,045               | 3,045              | 3,045               | 2,191                |
| Adjusted R-squared                     | 0.244             | 0.291             | 0.292               | 0.293              | 0.301               | 0.287                |

The dependent variable is Log (1+Births).  $\beta_X + \beta_{L.X} = 0, \text{pv}$  is the p-value of equality to zero of the sum of coefficients of the variable measured in t and in t+1. Period dummies and City FE always included. Standard errors (clustered at the NUTS 2 region level) in parentheses. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table 8: Population and Immigrants

|  | (1)               | (2)               | (3)                 | (4)                | (5)                 | (6)                  |
|--|-------------------|-------------------|---------------------|--------------------|---------------------|----------------------|
| Log (Population)                       | -0.008<br>(0.012) |                   |                     |                    |                     | -0.013<br>(0.013)    |
| L.Log (Population)                     | 0.016*<br>(0.009) |                   |                     |                    |                     | 0.011<br>(0.008)     |
| Bishop                                 |                   | -0.022<br>(0.018) |                     |                    |                     | -0.016<br>(0.023)    |
| L.Bishop                               |                   | -0.013<br>(0.018) |                     |                    |                     | -0.061***<br>(0.022) |
| Archbishop                             |                   |                   | 0.026<br>(0.030)    |                    |                     | 0.026<br>(0.032)     |
| L.Archbishop                           |                   |                   | -0.055**<br>(0.023) |                    |                     | -0.089***<br>(0.026) |
| Capital                                |                   |                   |                     | 0.055**<br>(0.027) |                     | 0.035<br>(0.024)     |
| L.Capital                              |                   |                   |                     | 0.055*<br>(0.030)  |                     | 0.051<br>(0.034)     |
| Commune                                |                   |                   |                     |                    | 0.037***<br>(0.013) | 0.034**<br>(0.015)   |
| L.Commune                              |                   |                   |                     |                    | 0.037***<br>(0.012) | 0.029**<br>(0.014)   |
| $\beta_X + \beta_{L.X} = 0, \text{pv}$ | 0.582             | 0.102             | 0.395               | 0.018              | 0.000               |                      |
| Observations                           | 3,434             | 3,045             | 3,045               | 3,045              | 3,045               | 2,191                |
| Adjusted R-squared                     | 0.141             | 0.138             | 0.138               | 0.144              | 0.146               | 0.147                |

The dependent variable is Log (1+Immigrants).  $\beta_X + \beta_{L.X} = 0, \text{pv}$  is the p-value of equality to zero of the sum of coefficients of the variable measured in t and in t+1. Period dummies and City FE always included. Standard errors (clustered at the NUTS 2 region level) in parentheses. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table 9: Famous Creatives and Wage of Skilled Workers

| Dependent Variable     | Y=Log(1+Un.Births.) |           | Y=Log(1+Un.Immig.) |           |
|------------------------|---------------------|-----------|--------------------|-----------|
|                        | Log(1+Un.Births.)   | Log(Wage) | Log(1+Un.Immig.)   | Log(Wage) |
| F5.Y                   |                     | 0.019*    |                    | 0.005     |
| F4.Y                   |                     | -0.002    |                    | 0.005     |
| F3.Y                   |                     | -0.006    |                    | 0.005     |
| F2.Y                   |                     | 0.001     |                    | -0.006    |
| F1.Y                   |                     | -0.014    |                    | -0.009    |
| Y                      |                     | 0.006     |                    | -0.009    |
| L1.Y                   | 0.322***            | -0.004    | 0.377***           | -0.002    |
| L2.Y                   | 0.175***            | 0.007     | 0.188***           | 0.023**   |
| L3.Y                   | 0.149**             | 0.014     | 0.085*             | -0.021    |
| L4.Y                   | 0.128**             | -0.001    | 0.121**            | 0.019*    |
| L5.Y                   | 0.078**             | 0.012     | 0.033              | 0.011     |
| Log(Wage)              | 0.093               |           | -0.181             |           |
| L1.Log(Wage)           | -0.348**            | 0.842***  | -0.011             | 0.853***  |
| L2.Log(Wage)           | 0.264*              | 0.052     | 0.065              | 0.049     |
| L3.Log(Wage)           | -0.001              | -0.011    | 0.193              | -0.011    |
| L4.Log(Wage)           | 0.134               | 0.016     | -0.064             | 0.021     |
| L5.Log(Wage)           | -0.126              | -0.018    | 0.058              | -0.020    |
| $F^{L.Wage}, pv$       | 0.880               | 0.000     | 0.528              | 0.000     |
| $F^{L.Births}, pv$     | 0.000               | 0.007     |                    |           |
| $F^{L.Immigrants}, pv$ |                     |           | 0.000              | 0.252     |
| $F^{F.Births}, pv$     |                     | 0.878     |                    |           |
| $F^{F.Immigrants}, pv$ |                     |           |                    | 0.976     |

The frequency of observation is a decade. The variable Wage represents the average nominal wage of skilled workers over the decade, measured in silver grams. The variable Un.Births is the (unscaled) number of Births.  $F^{L.X}, pv$  is the p-value of the F-test of joint significance of the lags of the variable X when X is the dependent variable, and of the contemporaneous and lag values when X is not the dependent variable.  $F^{F.X}, pv$  is the p-value of the F-test of joint significance of its leads. Period dummies always included.

Table 10: Commune and Births, OLS Regressions

|                                 | (1)                 | (2)                 | (3)                 | (4)                 |
|---------------------------------|---------------------|---------------------|---------------------|---------------------|
| Commune                         | 0.072***<br>(0.012) | 0.067***<br>(0.011) | 0.070***<br>(0.012) | 0.058***<br>(0.012) |
| Large state                     |                     | -0.033**<br>(0.014) | -0.035**<br>(0.014) | -0.008<br>(0.013)   |
| Bishop                          |                     | -0.032<br>(0.023)   | -0.028<br>(0.023)   | -0.009<br>(0.022)   |
| Archbishop                      |                     | -0.047<br>(0.032)   | -0.058*<br>(0.032)  | -0.056*<br>(0.030)  |
| Capital                         |                     | 0.057**<br>(0.025)  | 0.060**<br>(0.025)  | 0.064***<br>(0.025) |
| Plundered                       |                     | -0.019<br>(0.027)   | -0.016<br>(0.026)   | -0.015<br>(0.024)   |
| Log (Population)                |                     |                     | -0.005<br>(0.009)   | -0.014*<br>(0.008)  |
| University                      |                     |                     | 0.059**<br>(0.024)  | 0.059**<br>(0.023)  |
| Spatial Lag of Log (1 + Births) |                     |                     |                     | 1.016***<br>(0.141) |
| Observations                    | 3,045               | 3,045               | 3,045               | 3,045               |
| Adjusted R-squared              | 0.300               | 0.305               | 0.308               | 0.377               |

Dependent Variable is Log (1 + Births). Standard errors (clustered at the NUTS 2 region level) in parentheses. Period Dummies and City FE always included. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.



Table 11: IV Estimates of the Effect of Commune on Births

|                                 | (1)                 | (2)                 | (3)                 | (4)                 |
|---------------------------------|---------------------|---------------------|---------------------|---------------------|
| Panel A: 2SLS estimates         |                     |                     |                     |                     |
| Commune                         | 0.178***<br>(0.052) | 0.168***<br>(0.052) | 0.172***<br>(0.057) | 0.124**<br>(0.056)  |
| Spatial Lag of Log (1 + Births) |                     |                     |                     | 0.996***<br>(0.140) |
| Observations                    | 2,961               | 2,961               | 2,961               | 2,961               |
| Adjusted R-squared              | 0.116               | 0.127               | 0.129               | 0.225               |
| Fstat, instrum., 1st stage      | 22.67               | 20.40               | 18.26               | 18.10               |
| Baseline Controls               | NO                  | YES                 | YES                 | YES                 |
| Additional Controls             | NO                  | NO                  | YES                 | YES                 |
| Panel B: First stage estimates  |                     |                     |                     |                     |
| Regional Commune                | 1.508***<br>(0.316) | 1.498***<br>(0.331) | 1.446***<br>(0.338) | 1.437***<br>(0.336) |
| Panel C: Reduced form estimates |                     |                     |                     |                     |
| Regional Commune                | 0.256***<br>(0.061) | 0.234***<br>(0.061) | 0.231***<br>(0.063) | 0.163***<br>(0.058) |

Dependent Variable is Log (1 + Births). Standard errors (clustered at the NUTS 2 region level) in parentheses. Period Dummies and City FE always included. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table 12: Commune and Births: QMLE Poisson Estimates

|                                | (1)                 | (2)                 |
|--------------------------------|---------------------|---------------------|
| Commune                        | 0.905**<br>(0.386)  | 0.941***<br>(0.345) |
| Spatial Lag of Unscaled Births | 0.086***<br>(0.024) | 0.059***<br>(0.021) |
| Large state                    | 0.092<br>(0.132)    | 0.191<br>(0.116)    |
| Bishop                         | -0.771*<br>(0.445)  | -0.734<br>(0.466)   |
| Archbishop                     | -0.243<br>(0.468)   | -0.417<br>(0.549)   |
| Capital                        | 0.608**<br>(0.279)  | 0.626**<br>(0.288)  |
| Plundered                      | -0.340**<br>(0.172) | -0.265<br>(0.176)   |
| University                     | -0.260<br>(0.311)   | -0.179<br>(0.333)   |
| Observations                   | 2,012               | 2,012               |
| Log-likelihood                 | -2792               | -2759               |
| Population Decile Dummies      | YES                 | NO                  |
| Log Population Dummies         | NO                  | YES                 |

Dependent variable is number of births (not scaled by population). Estimation method: QMLE Poisson. Period dummies and city FE always included. In Col 1 we enter separate dummies for belonging to each of the deciles going from the first to the ninth, plus one dummy for belonging to the set of percentiles from the 90th to the 94th, one for the percentiles 95th to 98th; thus, the default group consists of cities belonging to the 99th percentile. In Col 2 we split the range of variation of Log(Population) in the entire sample into 10 equally sized intervals, and enter a dummy variable for each interval except the last one (which is the default). Standard errors (clustered at the city level) in parentheses.

\* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table 13: Determinants of the migration of famous creatives

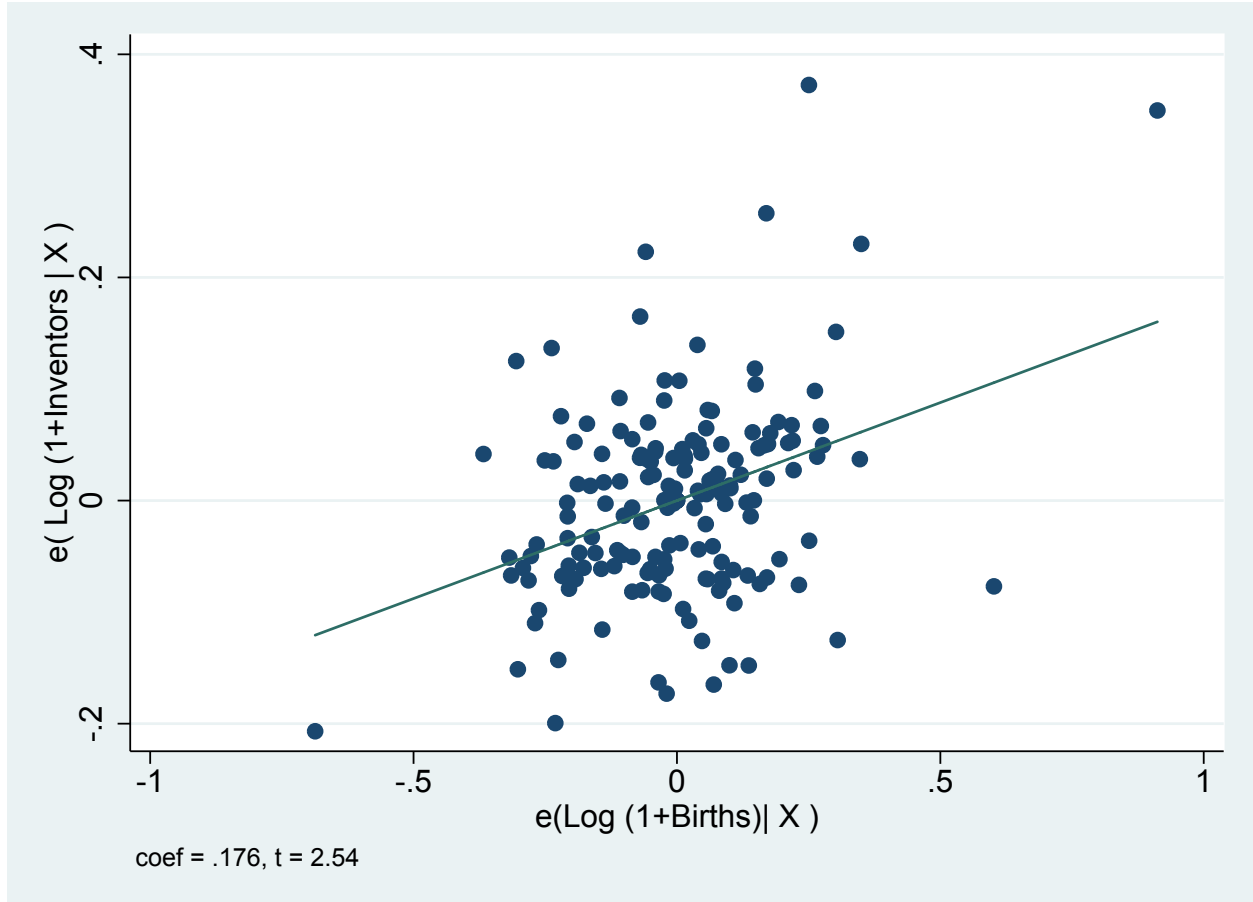
|                            | (1)                  | (2)                  | (3)                  |
|----------------------------|----------------------|----------------------|----------------------|
| To Population              | 0.209***<br>(0.056)  | -0.062***<br>(0.021) | -0.038<br>(0.026)    |
| To Commune                 | 0.959***<br>(0.157)  | 0.510**<br>(0.209)   | 0.520**<br>(0.213)   |
| To Plundered               | -0.666**<br>(0.281)  | 0.098<br>(0.195)     | 0.106<br>(0.215)     |
| To Capital                 | 2.846***<br>(0.299)  | 0.957***<br>(0.153)  | 0.857***<br>(0.173)  |
| To Archbishop              | 0.128<br>(0.257)     | -0.284<br>(0.268)    | -0.211<br>(0.309)    |
| To University              | 0.405*<br>(0.226)    | -0.428*<br>(0.227)   | -0.474<br>(0.289)    |
| To Bishop                  | -0.302<br>(0.215)    | -0.591*<br>(0.328)   | -0.538<br>(0.336)    |
| Log From (Unscaled) Births | 0.909***<br>(0.046)  | 0.950***<br>(0.037)  |                      |
| From Population            | -0.113***<br>(0.029) | 0.001<br>(0.022)     |                      |
| From Commune               | 0.091<br>(0.074)     | 0.095<br>(0.109)     |                      |
| From Plundered             | 0.316***<br>(0.091)  | 0.117<br>(0.086)     |                      |
| From Capital               | 0.071<br>(0.120)     | 0.035<br>(0.103)     |                      |
| From Archbishop            | 0.150<br>(0.122)     | 0.051<br>(0.363)     |                      |
| From University            | 0.007<br>(0.064)     | 0.169<br>(0.112)     |                      |
| From Bishop                | 0.022<br>(0.097)     | -0.385***<br>(0.129) |                      |
| Distance                   | -0.176***<br>(0.032) | -0.167***<br>(0.019) | -0.165***<br>(0.019) |
| Same First Lang            | 1.529***<br>(0.234)  | 1.379***<br>(0.112)  | 1.361***<br>(0.110)  |
| Same NUTS 1                | 0.537***<br>(0.147)  | 0.624***<br>(0.087)  | 0.594***<br>(0.087)  |
| Same State                 | 0.628***<br>(0.251)  | 1.486***<br>(0.107)  | 1.573***<br>(0.120)  |
| Dyadic Observations        | 714,855              | 714,855              | 714,855              |
| Period Dummies             | YES                  | YES                  | NO                   |
| Destination FE             | NO                   | YES                  | YES                  |
| Origin FE                  | NO                   | YES                  | NO                   |
| Origin-Year FE             | NO                   | NO                   | YES                  |

In this Table we estimate a gravity equation on dyadic data; estimation method is QMLE Poisson. Dependent Variable is Number of Immigrants. Twoway clustered standard errors (by origin and by destination) in parentheses. Population size measured in 100,000. Distance measured in 100 Km. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

# Appendix - For Online Publication

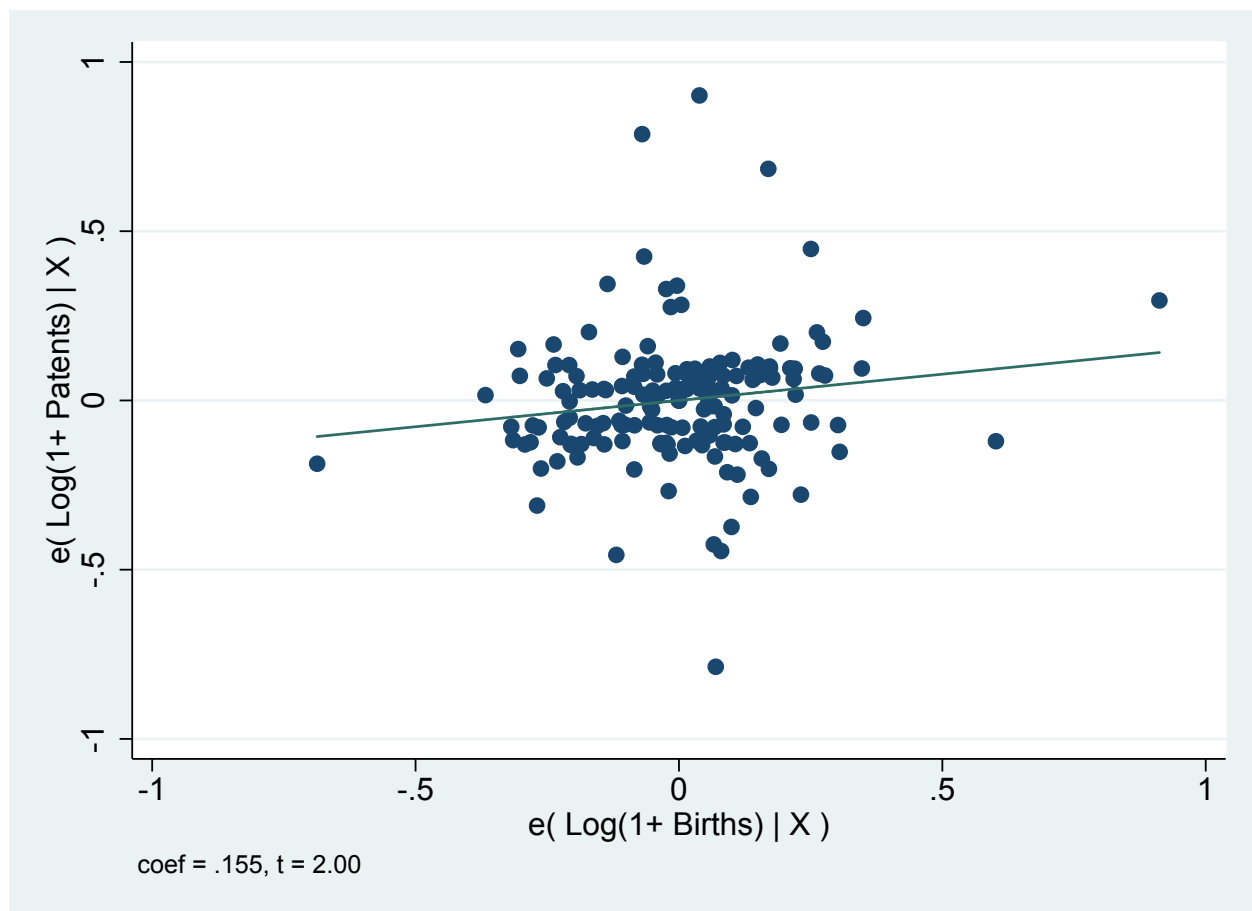
## A.I Additional Figure and Tables

Figure A.1: Famous Creatives and Inventors, added-variable plot



We control for City FE, Century FE, Large state, Bishop, Archbishop, Capital, Plundered, Commune, Population, University. Standard Errors clustered by Region.

Figure A.2: Famous Creatives and Patents, added-variable plot



We control for City FE, Century FE, Large state, Bishop, Archbishop, Capital, Plundered, Commune, Population, University. Standard Errors clustered by Region.

Figure A.3: Skilled Wages for 5 Prominent Cities

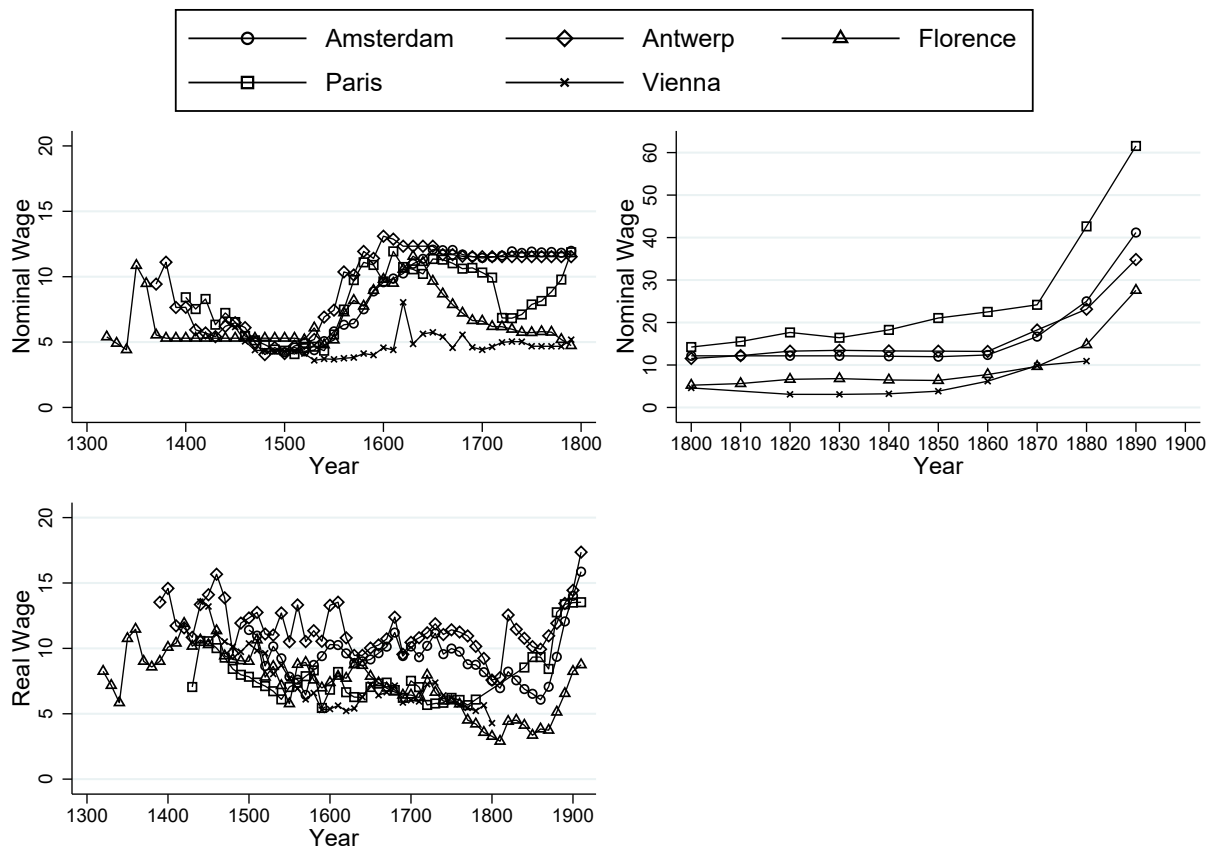
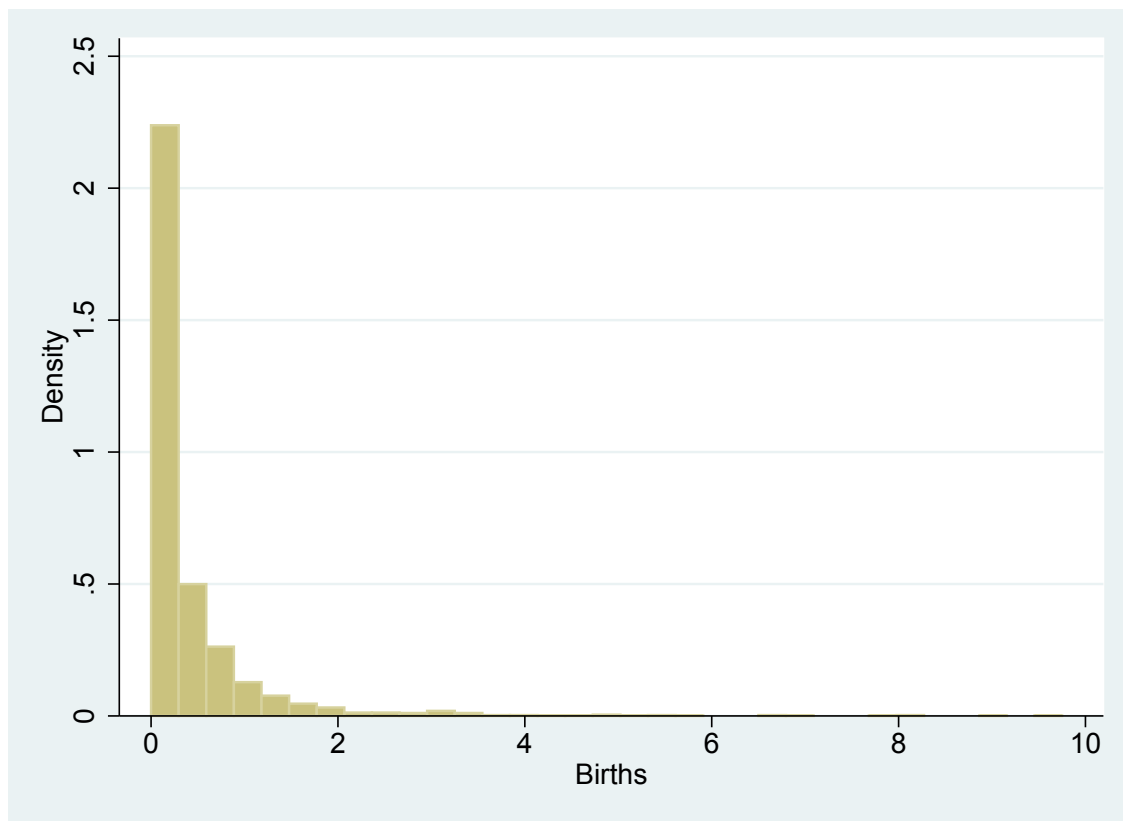
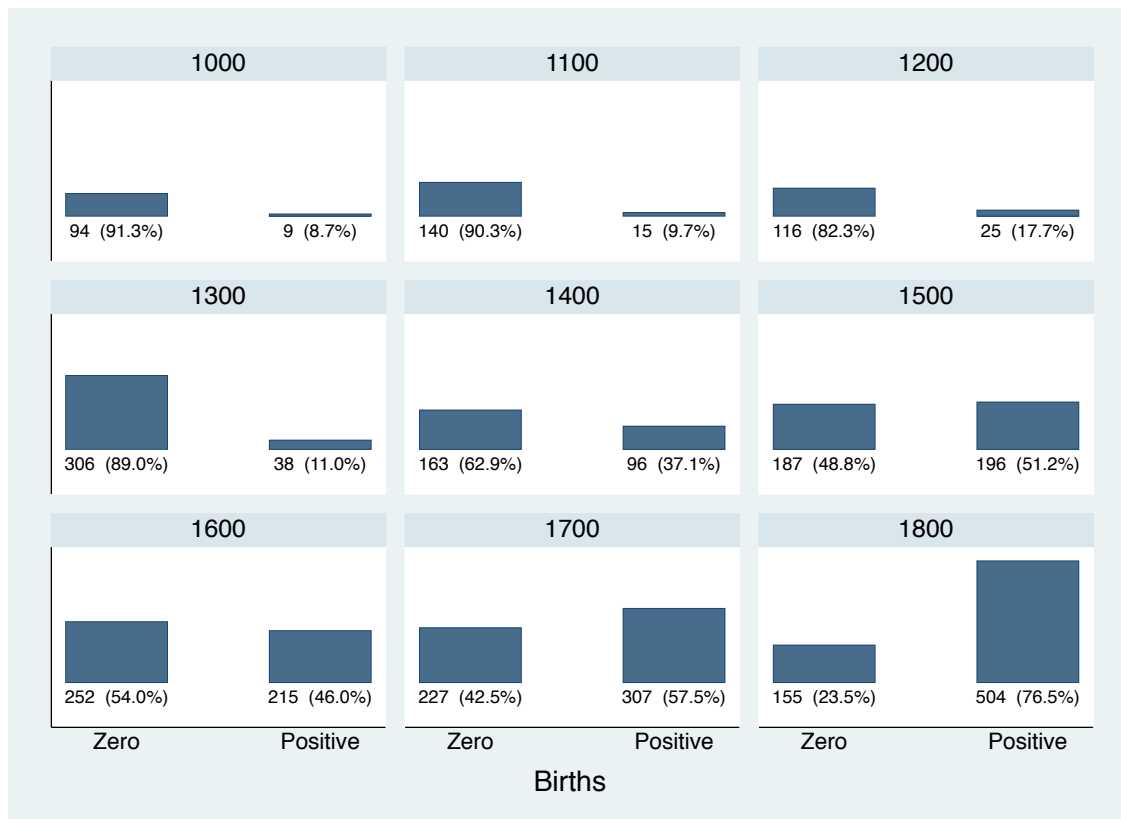


Figure A.4: Births, XIX century



This Figure displays the spatial distribution of Births in the XIX century, the end of our sample period.

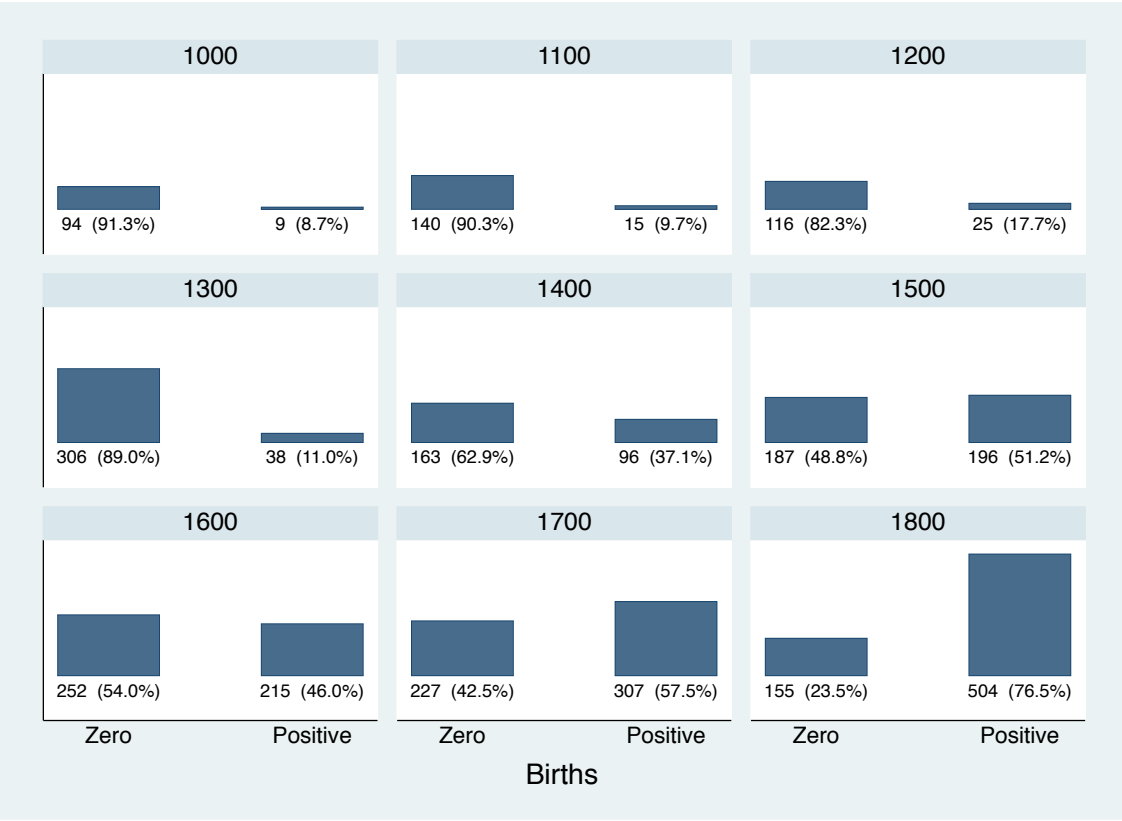
Figure A.5: Share of Cities with Zero and Positive Births, by Century



The Figure reports the share of cities by century with zero *Births*

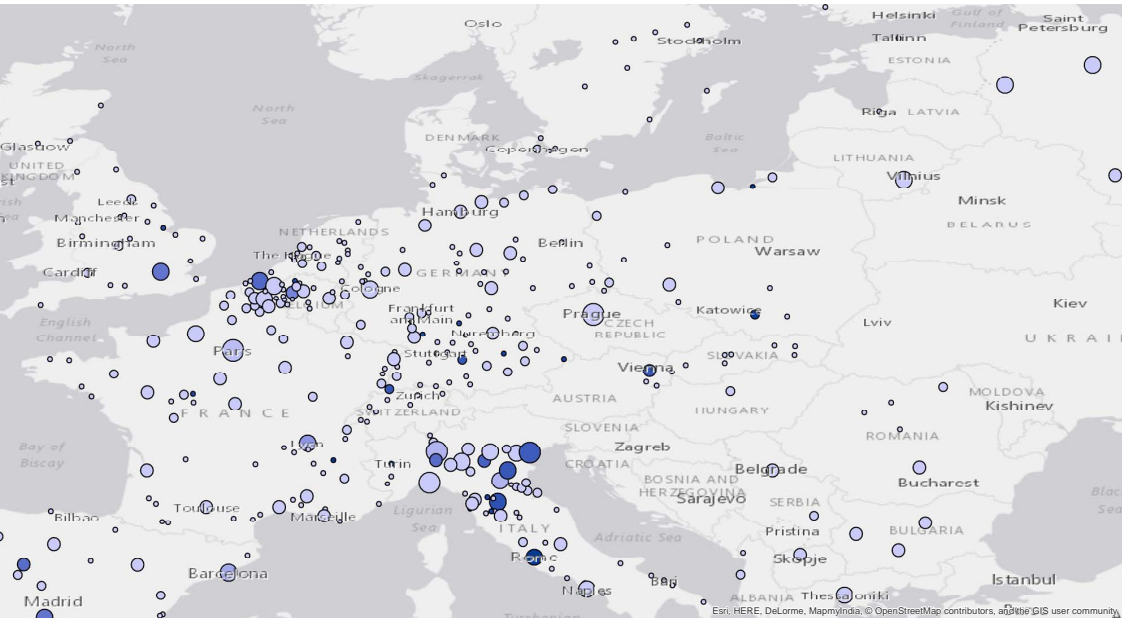


Figure A.6: Share of Cities with Zero and Positive Immigrants, by Century



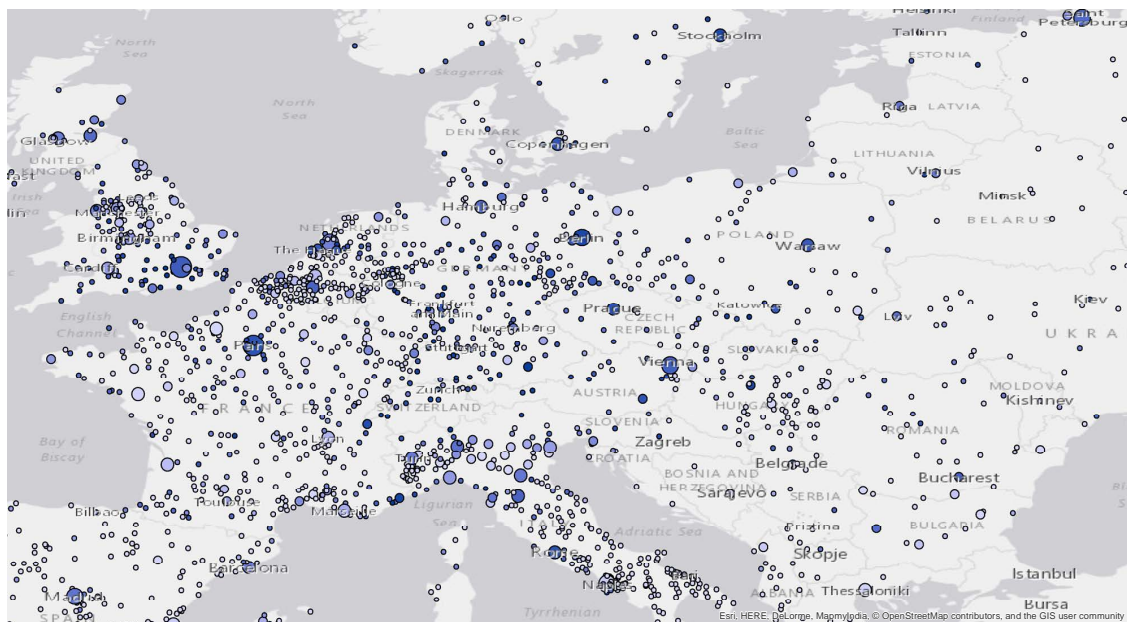
The Figure reports the share of cities by century with zero *Immigrants*

Figure A.7: Spatial Distribution of Immigrants, XVth century



the darker the tone, the higher the number of famous immigrants, per 1000 inhabitants. The larger the circle, the larger the population of the city. See Table A.9 for descriptive statistics.

Figure A.8: Spatial Distribution of Immigrants, XIXth century



The darker the tone, the higher the number of famous immigrants, per 1000 inhabitants. The larger the circle, the larger the population of the city. See Table A.10 for descriptive statistics and Figure 2 for further notes.

Figure A.9: Commune Status over Time

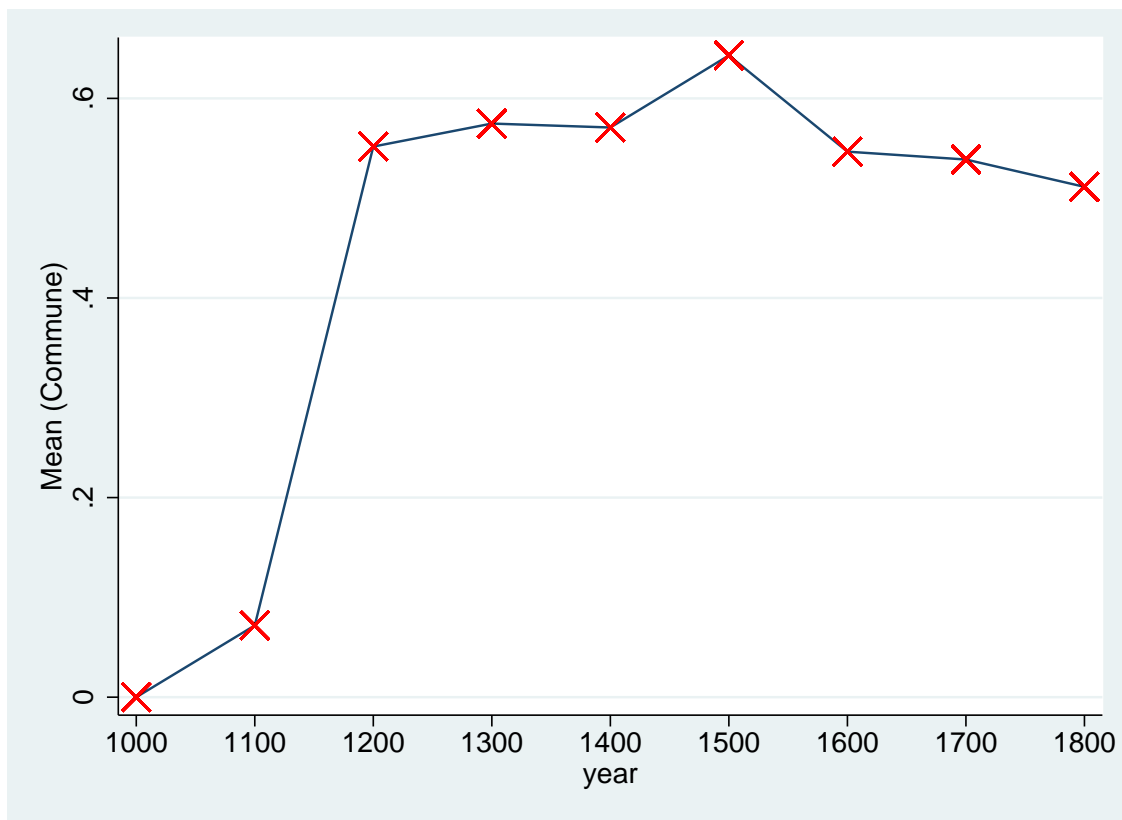


Figure A.10: Entry into Commune Status



Figure A.11: Exit from Commune Status

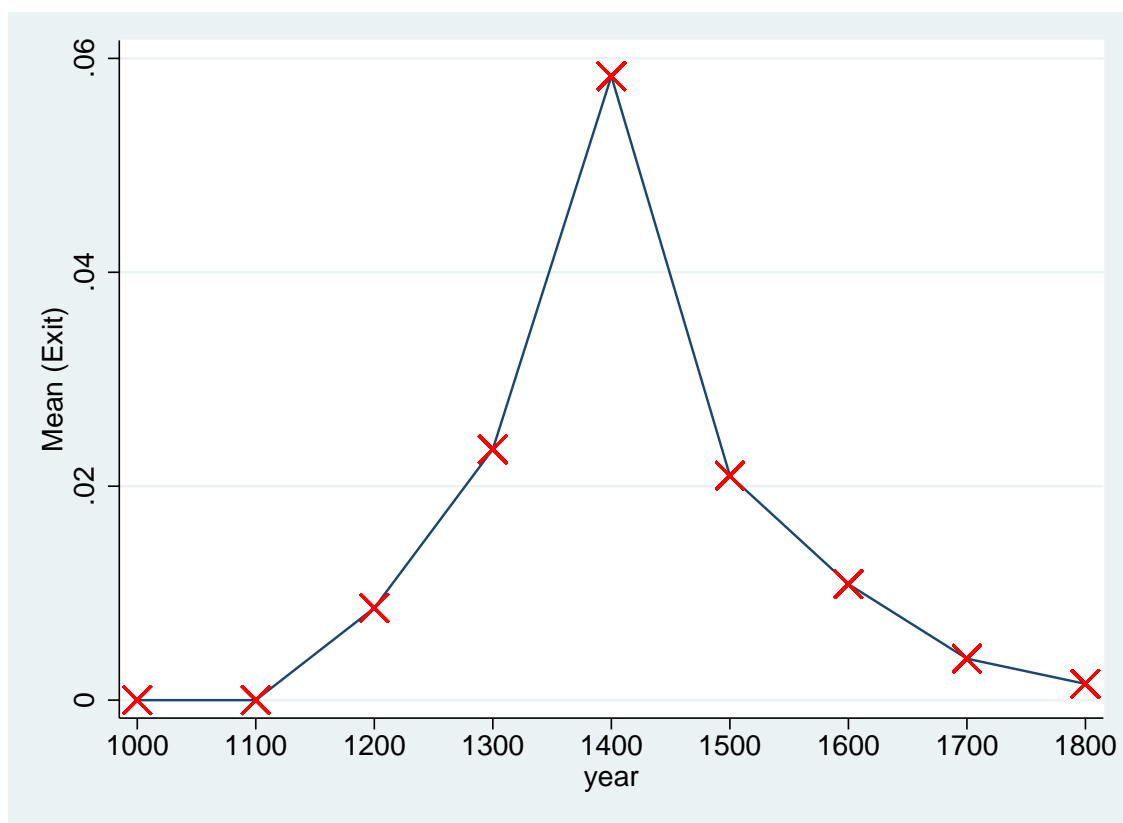


Table A.1: Polymaths

|   | Number of people |
|---|------------------|
| Performing arts + Non-performing arts                           | 1072             |
| Performing arts + Humanities and Sciences                       | 40               |
| Performing arts + Business                                      | 15               |
| Non-performing arts + Humanities and Sciences                   | 1245             |
| Non-performing arts + Business                                  | 211              |
| Humanities and Sciences + Business                              | 181              |
| Total two categories  | 2764             |
| Performing arts + Non-performing arts + Humanities and Sciences | 180              |
| Performing arts + Non-performing arts + Business                | 10               |
| Performing arts + Humanities and Sciences + Business            | 2                |
| Non-performing arts + Humanities and Sciences + Business        | 55               |
| Total three categories  | 247              |
| All four categories   | 5                |
| Total   | 21906            |

Note: Many of the creative individuals we considered achieved fame in multiple fields. For example, each of those listed under “Non-performing arts + Humanities and Sciences” achieved prominence both as a non-performing artist and in some field of the humanities or sciences, though not as a performing artist or in business.

Table A.2: Time Coverage of Data on Wages

| City       | Period    |
|------------|-----------|
| Amsterdam  | 1500-1910 |
| Antwerp    | 1373-1913 |
| Augsburg   | 1502-1803 |
| Barcelona  | 1500-1804 |
| Cambridge  | 1450-1700 |
| Canterbury | 1450-1700 |
| Cavaillon  | 1600-1785 |
| Dover      | 1450-1700 |
| Edinburgh  | 1553-1642 |
| Florence   | 1326-1913 |
| Gdansk     | 1535-1814 |
| Ghent      | 1835-1914 |
| Krakow     | 1409-1910 |
| Leipzig    | 1520-1913 |
| London     | 1264-1913 |
| Lviv       | 1520-1800 |
| Lyon       | 1500-1592 |
| Madrid     | 1520-1913 |
| Milan      | 1520-1913 |
| Munich     | 1427-1765 |
| Naples     | 1514-1806 |
| Oxford     | 1264-1913 |
| Paris      | 1400-1911 |
| Strasbourg | 1395-1875 |
| Valencia   | 1392-1785 |
| Valladolid | 1502-1560 |
| Vienna     | 1440-1913 |
| Warsaw     | 1558-1913 |

Note: Data do not necessarily cover the whole period as reported here; some gaps may be present.  
Sources: Allen (2001), Bennisar (1999), Boulton (1996), Cabourdin (1968), Feliu (1991), Gibson and Smout (1995), Rappaport (2002), and Scholliers and Avondts (1977)

Table A.3: Born in Florence (1100-1499)

| Name                                | Year of Birth | Year of Death | Place of Death | Occupation        |
|-------------------------------------|---------------|---------------|----------------|-------------------|
| Gaddo Gaddi                         | 1239          | 1312          | Florence       | Painter           |
| Cimabue                             | 1240          | 1302          | Pisa           | Painter           |
| Guido Cavalcanti                    | 1255          | 1300          | Florence       | Poet              |
| Dante Alighieri                     | 1265          | 1321          | Ravenna        | Poet              |
| Giotto di Bondone                   | 1267          | 1337          | Florence       | Painter           |
| Taddeo Gaddi                        | 1300          | 1366          | Florence       | Painter           |
| Antonio Pucci                       | 1310          | 1388          | Florence       | Poet              |
| Giotto                              | 1324          | 1357          | Florence       | Painter           |
| Baldassarre Bonaiuti                | 1336          | 1385          | Florence       | Historian         |
| Giovanni di Bicci de' Medici        | 1360          | 1429          | Florence       | Banker            |
| Palla Strozzi                       | 1372          | 1462          | Padova         | Banker            |
| Andrea Stefani                      | 1375          | 1460          | Lucca          | Composer          |
| Filippo Brunelleschi                | 1377          | 1446          | Florence       | Architect         |
| Lorenzo Ghiberti                    | 1378          | 1455          | Florence       | Sculptor          |
| Donatello                           | 1386          | 1466          | Florence       | Sculptor          |
| Cosimo de' Medici                   | 1389          | 1464          | Florence       | Banker            |
| Fra Angelico                        | 1395          | 1455          | Rome           | Painter           |
| Lorenzo il Vecchio                  | 1395          | 1440          | Florence       | Banker            |
| Michelozzo                          | 1396          | 1472          | Florence       | Architect         |
| Paolo Uccello                       | 1397          | 1475          | Florence       | Painter           |
| Paolo dal Pozzo Toscanelli          | 1397          | 1482          | Florence       | Astrologer        |
| Filarete                            | 1400          | 1469          | Rome           | Architect         |
| Luca della Robbia                   | 1400          | 1482          | Florence       | Sculptor          |
| Filippo Lippi                       | 1406          | 1469          | Spoletto       | Painter           |
| Matteo Palmieri                     | 1406          | 1475          | Florence       | Historian         |
| Bernardo Rossellino                 | 1409          | 1464          | Florence       | Sculptor          |
| Piero di Cosimo de' Medici          | 1416          | 1469          | Florence       | Banker and Patron |
| Agostino di Duccio                  | 1418          | 1481          | Perugia        | Sculptor          |
| Andrea del Castagno                 | 1421          | 1457          | Florence       | Painter           |
| Benozzo Gozzoli                     | 1421          | 1497          | Pistoia        | Painter           |
| Giovanni di Cosimo de' Medici       | 1421          | 1463          | Florence       | Banker            |
| Lucrezia Tornabuoni                 | 1425          | 1482          | Florence       | Writer            |
| Antonio Rossellino                  | 1427          | 1479          | Florence       | Sculptor          |
| Antonio Pollaiuolo                  | 1429          | 1498          | Rome           | Painter           |
| Pierfrancesco di Lorenzo de' Medici | 1430          | 1476          | Florence       | Banker            |
| Marsilio Ficino                     | 1433          | 1499          | Florence       | Philosopher       |
| Andrea del Verrocchio               | 1435          | 1488          | Venezia        | Painter           |
| Andrea della Robbia                 | 1435          | 1525          | Florence       | Sculptor          |
| Cosimo Rosselli                     | 1439          | 1507          | Florence       | Painter           |

|                                     |      |      |          |                     |
|-------------------------------------|------|------|----------|---------------------|
| Jacopo da Sellaio                   | 1441 | 1493 | Florence | Painter             |
| Piero Pollaiuolo                    | 1443 | 1496 | Rome     | Painter             |
| Sandro Botticelli                   | 1445 | 1510 | Florence | Painter             |
| Francesco Botticini                 | 1446 | 1498 | Florence | Painter             |
| Domenico Ghirlandaio                | 1449 | 1494 | Florence | Painter             |
| Lorenzo de' Medici                  | 1449 | 1492 | Florence | Writer              |
| Baccio Pontelli                     | 1450 | 1492 | Urbino   | Architect           |
| Antonio da Sangallo the Elder       | 1453 | 1534 | Florence | Architect           |
| Girolamo Benivieni                  | 1453 | 1542 | Florence | Poet                |
| Amerigo Vespucci                    | 1454 | 1512 | Sevilla  | Explorer            |
| Pietro Accolti                      | 1455 | 1532 | Rome     | Cardinal and Writer |
| Benedetto Buglioni                  | 1459 | 1521 | Florence | Sculptor            |
| Lorenzo di Credi                    | 1459 | 1537 | Florence | Painter             |
| Piero di Cosimo                     | 1462 | 1521 | Florence | Painter             |
| Lorenzo di Pierfrancesco de' Medici | 1463 | 1503 | Florence | Banker              |
| Niccolo Machiavelli                 | 1469 | 1527 | Florence | Historian           |
| Pietro Torrigiano                   | 1472 | 1528 | Sevilla  | Sculptor            |
| Giovanni Francesco Rustici          | 1474 | 1554 | Tours    | Painter             |
| Mariotto Albertinelli               | 1474 | 1515 | Florence | Painter             |
| Giovanni Rucellai                   | 1475 | 1525 | Rome     | Poet                |
| Jacopo Nardi                        | 1476 | 1563 | Venezia  | Historian           |
| Franciabigio                        | 1482 | 1525 | Florence | Painter             |
| Francesco Guicciardini              | 1483 | 1540 | Florence | Historian           |
| Ridolfo Ghirlandajo                 | 1483 | 1561 | Florence | Painter             |
| Antonio da Sangallo the Younger     | 1484 | 1546 | Terni    | Architect           |
| Jacopo Sansovino                    | 1486 | 1570 | Venezia  | Sculptor            |
| Girolamo della Robbia               | 1488 | 1566 | Paris    | Potter              |
| Pietro Aron                         | 1489 | 1550 | Florence | Composer            |
| Lorenzetto                          | 1490 | 1541 | Rome     | Sculptor            |
| Agnolo Firenzuola                   | 1493 | 1543 | Prato    | Poet                |
| Rosso Fiorentino                    | 1494 | 1540 | Paris    | Painter             |
| Benedetto Accolti jr                | 1497 | 1549 | Florence | Cardinal and Writer |

Note: Source is Freebase. If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample, within a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution).



Table A.4: Born in Amsterdam (1200-1699)

| Name                          | Year of Birth | Year of Death | Place of Death | Occupation    |
|-------------------------------|---------------|---------------|----------------|---------------|
| Pieter Aertsen                | 1508          | 1575          | Amsterdam      | Painter       |
| Lambert Sustris               | 1515          | 1584          | Venezia        | Painter       |
| Dirck Volckertszoon Coornhert | 1522          | 1590          | Gouda          | Writer        |
| Hendrik Laurenszoon Spiegel   | 1549          | 1612          | Alkmaar        | Writer        |
| Jan Pieterszoon Sweelinck     | 1562          | 1621          | Amsterdam      | Composer      |
| Jacob van Heemskerck          | 1567          | 1607          | Gibraltar      | Explorer      |
| Pieter Corneliszoon Hooft     | 1581          | 1647          | Den Haag       | Historian     |
| Laurens Reael                 | 1583          | 1637          | Amsterdam      | Admiral       |
| Simon Episcopius              | 1583          | 1643          | Amsterdam      | Theologian    |
| Gerbrand Adriaensz Bredero    | 1585          | 1618          | Amsterdam      | Poet          |
| Hendrick Avercamp             | 1585          | 1634          | Kampen         | Painter       |
| Andries Bicker                | 1586          | 1652          | Amsterdam      | Merchant      |
| Esaïas van de Velde           | 1587          | 1630          | Den Haag       | Painter       |
| Nicolaes Tulp                 | 1593          | 1674          | Den Haag       | Surgeon       |
| Pauwels van Hillegaert        | 1596          | 1640          | Amsterdam      | Painter       |
| Thomas de Keyser              | 1596          | 1667          | Amsterdam      | Painter       |
| Salomon de Bray               | 1597          | 1664          | Haarlem        | Architect     |
| Isaac Commelin                | 1598          | 1676          | Amsterdam      | Historian     |
| Michael van Langren           | 1598          | 1675          | Bruxelles      | Astronomer    |
| Cornelis de Graeff            | 1599          | 1664          | Amsterdam      | Merchant      |
| Pieter Codde                  | 1599          | 1678          | Amsterdam      | Painter       |
| Philip Vingboons              | 1607          | 1678          | Amsterdam      | Architect     |
| Salomon Koninck               | 1609          | 1656          | Amsterdam      | Painter       |
| Jan Asselijn                  | 1610          | 1652          | Amsterdam      | Painter       |
| Andries de Graeff             | 1611          | 1678          | Amsterdam      | Merchant      |
| Philip de Koninck             | 1619          | 1688          | Amsterdam      | Painter       |
| Gerbrand van den Eeckhout     | 1621          | 1674          | Amsterdam      | Painter       |
| Jan Baptist Weenix            | 1621          | 1660          | Utrecht        | Painter       |
| Jan Abrahamsz Beerstraten     | 1622          | 1666          | Amsterdam      | Painter       |
| Reinier Nooms                 | 1623          | 1667          | Amsterdam      | Painter       |
| Lambert Doomer                | 1624          | 1700          | Amsterdam      | Painter       |
| Willem Schellinks             | 1627          | 1678          | Amsterdam      | Painter       |
| Jan Hackaert                  | 1628          | 1685          | Amsterdam      | Painter       |
| Jan de Bisschop               | 1628          | 1671          | Den Haag       | Painter       |
| Johann van Waveren Hudde      | 1628          | 1704          | Amsterdam      | Mathematician |
| Adriaan Koerbagh              | 1632          | 1669          | Amsterdam      | Philosopher   |
| Baruch Spinoza                | 1632          | 1677          | Den Haag       | Philosopher   |
| Willem Drost                  | 1633          | 1659          | Venezia        | Painter       |
| Frederik de Moucheron         | 1633          | 1686          | Amsterdam      | Painter       |

|                         |      |      |             |                                |
|-------------------------|------|------|-------------|--------------------------------|
| Adriaen Backer          | 1635 | 1684 | Amsterdam   | Painter                        |
| Olfert Dapper           | 1635 | 1689 | Amsterdam   | Writer and Physician           |
| Jan Swammerdam          | 1637 | 1680 | Amsterdam   | Biologist                      |
| Meindert Hobbema        | 1638 | 1709 | Amsterdam   | Painter                        |
| Jan Weenix              | 1640 | 1719 | Amsterdam   | Painter                        |
| Karel Dujardin          | 1640 | 1678 | Venezia     | Painter                        |
| Johann Ludwig Hannemann | 1640 | 1724 | Kiel        | Chemist                        |
| Nicolaes Witsen         | 1641 | 1717 | Amsterdam   | Writer and Diplomat            |
| Burchard de Volder      | 1643 | 1709 | Leiden      | Philosopher                    |
| Abraham Storck          | 1644 | 1708 | Amsterdam   | Painter                        |
| Albert Meijeringh       | 1645 | 1714 | Amsterdam   | Painter                        |
| Romeyn de Hooghe        | 1645 | 1708 | Haarlem     | Sculptor                       |
| Nicolaes de Vree        | 1645 | 1702 | Alkmaar     | Painter                        |
| Johannes Voorhout       | 1647 | 1723 | Amsterdam   | Painter                        |
| Petrus Houttuyn         | 1648 | 1709 | Leiden      | Botanist                       |
| Govert Bidloo           | 1649 | 1713 | Leiden      | Poet, Physician and Playwright |
| Johannes Verkolje       | 1650 | 1693 | Delft       | Painter                        |
| Johannes van der Bent   | 1650 | 1690 | Amsterdam   | Painter                        |
| Jan Griffier            | 1652 | 1718 | London      | Painter                        |
| Jan Hoogsaat            | 1654 | 1730 | Amsterdam   | Painter                        |
| Dirk Dalens             | 1657 | 1687 | Amsterdam   | Painter                        |
| Johannes Schenck        | 1660 | 1712 | Duesseldorf | Composer                       |
| Abraham Alewijn         | 1664 | 1721 | Jakarta     | Poet and Jurist                |
| Rachel Ruysch           | 1664 | 1750 | Amsterdam   | Painter                        |
| Caspar Commelijn        | 1668 | 1731 | Amsterdam   | Botanist                       |
| Cornelis de Graeff II.  | 1671 | 1719 | Monnikendam | Banker                         |
| Jan van Huysum          | 1682 | 1749 | Amsterdam   | Painter                        |

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Note: Source is Freebase. If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample, within a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution).

Table A.5: Born in Antwerp (1200-1599)

| Name                       | Year of Birth | Year of Death | Place of Death | Occupation   |
|----------------------------|---------------|---------------|----------------|--------------|
| Jacobus Barbireau          | 1455          | 1491          | Antwerp        | Composer     |
| Anna Bijns                 | 1493          | 1575          | Antwerp        | Writer       |
| Jan Sanders van Hemessen   | 1500          | 1566          | Haarlem        | Painter      |
| Hieronymus Cock            | 1510          | 1570          | Antwerp        | Painter      |
| Cornelis Floris de Vriendt | 1514          | 1575          | Antwerp        | Sculptor     |
| Hubert Waelrant            | 1517          | 1595          | Antwerp        | Composer     |
| Abraham Ortelius           | 1527          | 1598          | Antwerp        | Cartographer |
| Matthew Wesenbeck          | 1531          | 1586          | Wittenberg     | Jurist       |
| Denis Calvaert             | 1540          | 1619          | Bologna        | Painter      |
| Joris Hoefnagel            | 1542          | 1601          | Vienna         | Painter      |
| Gillis van Coninxloo       | 1544          | 1607          | Amsterdam      | Painter      |
| Bartholomeus Spranger      | 1546          | 1611          | Prag           | Painter      |
| Martin Delrio              | 1551          | 1608          | Leuven         | Theologian   |
| Leonardus Lessius          | 1554          | 1623          | Leuven         | Theologian   |
| Hans Jordaens              | 1555          | 1630          | Delft          | Painter      |
| Jan de Wael I              | 1558          | 1633          | Antwerp        | Painter      |
| Jan Gruter                 | 1560          | 1627          | Heidelberg     | Philologist  |
| Joos de Momper             | 1564          | 1635          | Antwerp        | Painter      |
| Jacob de Gheyn II          | 1565          | 1629          | Den Haag       | Painter      |
| Abraham Janssens           | 1567          | 1632          | Antwerp        | Painter      |
| Sebald de Weert            | 1567          | 1603          | Sri Lanka      | Explorer     |
| Joris van Spilbergen       | 1568          | 1620          | Bergen-Op-Zoom | Explorer     |
| Frans Pourbus the younger  | 1569          | 1622          | Paris          | Painter      |
| Ambrosius Bosschaert       | 1573          | 1621          | Den Haag       | Painter      |
| Jacobus Boonen             | 1573          | 1655          | Bruxelles      | Bishop       |
| Sebastian Vrancx           | 1573          | 1647          | Antwerp        | Painter      |
| Hendrick van Balen         | 1575          | 1632          | Antwerp        | Painter      |
| Pieter Neeffs I            | 1578          | 1656          | Antwerp        | Painter      |
| Frans Hals                 | 1580          | 1666          | Haarlem        | Painter      |
| Artus Wolffort             | 1581          | 1641          | Antwerp        | Painter      |
| Caspar de Crayer           | 1582          | 1669          | Ghent          | Painter      |
| David Teniers the Elder    | 1582          | 1649          | Antwerp        | Painter      |
| Jacques l'Hermite          | 1582          | 1624          | Callao         | Merchant     |
| Caspar Barlaeus            | 1584          | 1648          | Amsterdam      | Poet         |
| Willem van Nieulandt II    | 1584          | 1635          | Amsterdam      | Painter      |
| Gijsbrecht Leytens         | 1586          | 1656          | Antwerp        | Painter      |
| Jan Wildens                | 1586          | 1653          | Antwerp        | Painter      |
| Andries van Eertvelt       | 1590          | 1652          | Antwerp        | Painter      |
| Daniel Seghers             | 1590          | 1661          | Antwerp        | Painter      |

|                      |      |      |           |          |
|----------------------|------|------|-----------|----------|
| Francisco Pelsaert   | 1590 | 1630 | Jakarta   | Merchant |
| Lucas de Wael        | 1591 | 1661 | Antwerp   | Painter  |
| Cornelis de Wael     | 1592 | 1667 | Rome      | Painter  |
| Jacob Jordaens       | 1593 | 1678 | Antwerp   | Painter  |
| Dirk van Hoogstraten | 1596 | 1640 | Dordrecht | Painter  |
| Jacob van Es         | 1596 | 1666 | Antwerp   | Painter  |
| Cornelis Schut       | 1597 | 1655 | Antwerp   | Painter  |
| Justus Sustermans    | 1597 | 1681 | Florence  | Painter  |
| Pieter Claesz        | 1597 | 1660 | Haarlem   | Painter  |
| Adriaen van Utrecht  | 1599 | 1652 | Antwerp   | Painter  |
| Anthony van Dyck     | 1599 | 1641 | London    | Painter  |

Note: Souce is Freebase. If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample, within a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution).

Table A.6: Born in Paris (1700-1899)

| <b>Name</b>   | <b>Year of Birth</b> | <b>Occupation</b> |
|---|----------------------|-------------------|
| Francois Boucher                                    | 1703                 | Painter           |
| Emilie du Chatelet                                  | 1706                 | Mathematician     |
| Alexis Clairault                                    | 1713                 | Mathematician     |
| Claude-Adrien Helvetius                             | 1715                 | Philosopher       |
| Jean le Rond d'Alembert                             | 1717                 | Mathematician     |
| Anne Robert Jacques Turgot, Baron de Laune          | 1727                 | Economist         |
| Louis Antoine de Bougainville                       | 1729                 | Explorer          |
| Pierre de Beaumarchais                              | 1732                 | Writer            |
| Donatien-Alphonse-Francois de Sade, Marquis de Sade | 1740                 | Writer            |
| Antoine Lavoisier                                   | 1743                 | Chemist           |
| Jacques-Louis David                                 | 1748                 | Painter           |
| Adrien-Marie Legendre                               | 1752                 | Mathematician     |
| Elisabeth Louise Vigee Le Brun                      | 1755                 | Artist            |
| Claude Henri de Rouvroy, comte de Saint-Simon       | 1760                 | Philosopher       |
| Anne Louise Germaine de Stael                       | 1766                 | Writer            |
| Antoine-Jean Gros                                   | 1771                 | Artist            |
| Jean-Baptiste Biot                                  | 1774                 | Physicist         |
| Sophie Germain                                      | 1776                 | Mathematician     |
| Augustin Louis Cauchy                               | 1789                 | Mathematician     |
| Gaspard-Gustave Coriolis                            | 1792                 | Physicist         |
| Jean-Baptiste Camille Corot                         | 1796                 | Painter           |
| Nicolas Leonard Sadi Carnot                         | 1796                 | Engineer          |
| Jules Michelet                                      | 1798                 | Writer            |
| Adolphe-Charles Adam                                | 1803                 | Composer          |
| Charles Lucien Bonaparte                            | 1803                 | Biologist         |
| Prosper Merimee                                     | 1803                 | Writer            |
| George Sand   | 1804                 | Writer            |
| Alexis de Tocqueville                               | 1805                 | Historian         |
| Gerard de Nerval                                    | 1808                 | Writer            |
| Baron Haussmann                                     | 1809                 | Architect         |
| Alfred de Musset                                    | 1810                 | Writer            |
| Eliphas Levi  | 1810                 | Writer            |
| Theodore Rousseau                                   | 1812                 | Painter           |
| Charles-Valentin Alkan                              | 1813                 | Musician          |
| Eugene Viollet-le-Duc                               | 1814                 | Architect         |
| Charles Gounod                                      | 1818                 | Composer          |
| Hippolyte Fizeau                                    | 1819                 | Physicist         |
| Leon Foucault                                       | 1819                 | Physicist         |
| Nadar   | 1820                 | Photographer      |

|                                |      |               |
|--------------------------------|------|---------------|
| Charles Baudelaire             | 1821 | Writer        |
| Frederic Passy                 | 1822 | Economist     |
| Alexandre Dumas                | 1824 | Writer        |
| Pierre Jules Cesar Janssen     | 1824 | Astronomer    |
| Jean-Martin Charcot            | 1825 | Physician     |
| Gustave Moreau                 | 1826 | Artist        |
| Marcellin Berthelot            | 1827 | Chemist       |
| Edouard Manet                  | 1832 | Painter       |
| Edgar Degas                    | 1834 | Painter       |
| Camille Saint-Saens            | 1835 | Composer      |
| Georges Bizet                  | 1838 | Composer      |
| Alfred Sisley                  | 1839 | Artist        |
| Sully Prudhomme                | 1839 | Writer        |
| Emile Zola                     | 1840 | Writer        |
| Auguste Rodin                  | 1840 | Sculptor      |
| Claude Monet                   | 1840 | Painter       |
| Stephane Mallarme              | 1842 | Writer        |
| Anatole France                 | 1844 | Writer        |
| Sarah Bernhardt                | 1844 | Actor         |
| Charles Louis Alphonse Laveran | 1845 | Physician     |
| Gustave Caillebotte            | 1848 | Painter       |
| Joris-Karl Huysmans            | 1848 | Writer        |
| Paul Gauguin                   | 1848 | Painter       |
| Vilfredo Pareto                | 1848 | Economist     |
| Charles Robert Richet          | 1850 | Biologist     |
| Henri Louis Le Chatelier       | 1850 | Chemist       |
| Antoine Henri Becquerel        | 1852 | Physicist     |
| Henri Moissan                  | 1852 | Chemist       |
| Rudolf Diesel                  | 1858 | Inventor      |
| Georges-Pierre Seurat          | 1859 | Artist        |
| Henri Bergson                  | 1859 | Philosopher   |
| Pierre Curie                   | 1859 | Physicist     |
| Georges Melies                 | 1861 | Film Director |
| Paul Signac                    | 1863 | Painter       |
| Pierre de Coubertin            | 1863 | Historian     |
| Paul Dukas                     | 1865 | Composer      |
| Gaston Leroux                  | 1868 | Writer        |
| Andre Gide                     | 1869 | Writer        |
| Paul Langevin                  | 1872 | Physicist     |
| W. Somerset Maugham            | 1874 | Writer        |
| Maurice de Vlaminck            | 1876 | Artist        |
| Louis Renault                  | 1877 | Inventor      |
| Andre Citroen                  | 1878 | Engineer      |

|                         |      |             |
|-------------------------|------|-------------|
| Francis Picabia         | 1879 | Painter     |
| Roger Martin du Gard    | 1881 | Writer      |
| Jacques Maritain        | 1882 | Philosopher |
| Maurice Utrillo         | 1883 | Artist      |
| Robert Delaunay         | 1885 | Painter     |
| Nadia Boulanger         | 1887 | Musician    |
| Maurice Chevalier       | 1888 | Actor       |
| Gabriel Marcel          | 1889 | Philosopher |
| Jacques Ibert           | 1890 | Composer    |
| Marcel Dassault         | 1892 | Engineer    |
| Jean Renoir             | 1894 | Actor       |
| Andre Frederic Cournand | 1895 | Physician   |
| Basil Liddell Hart      | 1895 | Writer      |
| Frederic Joliot-Curie   | 1897 | Chemist     |
| Louis Aragon            | 1897 | Writer      |
| Georges Dumezil         | 1898 | Linguist    |
| Francis Poulenc         | 1899 | Composer    |

Note: Source is Yu et al., (2016). If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample, within a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution).

Table A.7: Born in Vienna (1800-1899)

| <b>Name</b>             | <b>Year of Birth</b> | <b>Occupation</b> |
|-------------------------|----------------------|-------------------|
| Johann Strauss I        | 1804                 | Composer          |
| Johann Strauss II       | 1825                 | Composer          |
| Otto Wagner             | 1841                 | Architect         |
| Ludwig Boltzmann        | 1844                 | Physicist         |
| Arthur Schnitzler       | 1862                 | Writer            |
| Gustav Klimt            | 1862                 | Painter           |
| Richard Adolf Zsigmondy | 1865                 | Chemist           |
| Gustav Meyrink          | 1868                 | Writer            |
| Karl Landsteiner        | 1868                 | Biologist         |
| Alfred Adler            | 1870                 | Psychologist      |
| Arnold Schoenberg       | 1874                 | Composer          |
| Hugo von Hofmannsthal   | 1874                 | Writer            |
| Fritz Kreisler          | 1875                 | Musician          |
| Robert Barany           | 1876                 | Physician         |
| Lise Meitner            | 1878                 | Physicist         |
| Martin Buber            | 1878                 | Philosopher       |
| Otto Weininger          | 1880                 | Philosopher       |
| Stefan Zweig            | 1881                 | Writer            |
| Melanie Klein           | 1882                 | Psychologist      |
| Anton Webern            | 1883                 | Composer          |
| Alban Berg              | 1885                 | Composer          |
| Erich von Stroheim      | 1885                 | Film Director     |
| Karl von Frisch         | 1886                 | Biologist         |
| Erwin Schrodinger       | 1887                 | Physicist         |
| Ludwig Wittgenstein     | 1889                 | Philosopher       |
| Egon Schiele            | 1890                 | Painter           |
| Fritz Lang              | 1890                 | Film Director     |
| Anna Freud              | 1895                 | Psychologist      |
| Friedrich Hayek         | 1899                 | Economist         |

Note: Source is Yu et al., (2016). If an individual is born or dies in a small city not included in the Bairoch et al. (1988) sample, we assign it to the closest city in the sample, within a threshold of 71 Km (corresponding to the 95th percentile of the distance distribution).



Table A.8: Correlation between Births, Immigrants, Population and Commune

|            | Births | Immigrants | Population | Commune |
|------------|--------|------------|------------|---------|
| Births     | 1      |            |            |         |
| Immigrants | 0.56   | 1          |            |         |
| Population | -0.03  | 0          | 1          |         |
| Commune    | 0.13   | 0.11       | 0.1        | 1       |

Table A.9: Summary Statistics, XVth century

| <b>Variable</b> | <b>Mean</b> | <b>(Std. Dev.)</b> | <b>Min.</b> | <b>Max.</b> | <b>N</b> |
|-----------------|-------------|--------------------|-------------|-------------|----------|
| Population      | 11615.752   | (17656.554)        | 1000        | 200000      | 419      |
| Births          | 0.094       | (0.251)            | 0           | 2           | 419      |
| Immigrants      | 0.053       | (0.157)            | 0           | 1.333       | 419      |

Table A.10: Summary Statistics, XIXth century

| <b>Variable</b> | <b>Mean</b> | <b>(Std. Dev.)</b> | <b>Min.</b> | <b>Max.</b> | <b>N</b> |
|-----------------|-------------|--------------------|-------------|-------------|----------|
| Population      | 12035.951   | (31100.963)        | 1000        | 948000      | 2114     |
| Births          | 0.617       | (1.168)            | 0           | 13.25       | 2114     |
| Immigrants      | 0.371       | (1.51)             | 0           | 30          | 2114     |

Table A.11: Commune, First Stage

|                                 | (1)                 | (2)                  | (3)                 | (4)                 | (5)                  |
|---------------------------------|---------------------|----------------------|---------------------|---------------------|----------------------|
| Regional Commune                | 1.450***<br>(0.207) | 1.381***<br>(0.205)  | 1.446***<br>(0.338) | 1.434***<br>(0.337) | 1.427***<br>(0.336)  |
| Large state                     |                     | -0.053***<br>(0.019) | -0.020<br>(0.032)   | -0.015<br>(0.032)   | -0.014<br>(0.032)    |
| Bishop                          |                     | 0.126***<br>(0.037)  | -0.016<br>(0.054)   | -0.012<br>(0.054)   | -0.012<br>(0.054)    |
| Archbishop                      |                     | 0.077<br>(0.048)     | -0.109*<br>(0.065)  | -0.107<br>(0.065)   | -0.103<br>(0.065)    |
| Capital                         |                     | 0.080<br>(0.050)     | -0.005<br>(0.055)   | -0.004<br>(0.055)   | -0.006<br>(0.054)    |
| Plundered                       |                     | -0.032<br>(0.024)    | -0.025<br>(0.033)   | -0.024<br>(0.033)   | -0.026<br>(0.033)    |
| Log (Population)                |                     |                      | 0.051***<br>(0.012) | 0.049***<br>(0.012) | 0.050***<br>(0.012)  |
| University                      |                     |                      | -0.012<br>(0.054)   | -0.012<br>(0.054)   | -0.017<br>(0.054)    |
| Spatial Lag of Log (1 + Births) |                     |                      |                     | 0.185**<br>(0.094)  | 0.226**<br>(0.093)   |
| Spatial Lag of Log Population   |                     |                      |                     |                     | -0.009***<br>(0.003) |
| Observations                    | 7,227               | 7,227                | 3,110               | 3,110               | 3,110                |
| Number of ID                    | 657                 | 657                  | 657                 | 657                 | 657                  |
| Adjusted R-squared              | 0.446               | 0.454                | 0.427               | 0.427               | 0.429                |

Dependent Variable is Commune. Standard errors (clustered at the NUTS 2 region level) in parentheses.  
 Period Dummies and City FE always included. \* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table A.12: IV Estimates of the Effect of Commune on Births, using the Yu et al. data

|                                 | (1)                | (2)                | (3)                | (4)                | (5)                  |
|---------------------------------|--------------------|--------------------|--------------------|--------------------|----------------------|
| Commune                         | 0.229**<br>(0.113) | 0.259**<br>(0.116) | 0.260**<br>(0.123) | 0.237*<br>(0.122)  | 0.228*<br>(0.122)    |
| Spatial Lag of Log (1 + Births) |                    |                    |                    | 0.473**<br>(0.209) | 0.538***<br>(0.206)  |
| Spatial Lag of Log Population   |                    |                    |                    |                    | -0.019***<br>(0.007) |
| Observations                    | 2,961              | 2,961              | 2,961              | 2,961              | 2,961                |
| Adjusted R-squared              | -0.171             | -0.170             | -0.171             | -0.164             | -0.160               |
| Fstat, instrum., 1st stage      | 21.60              | 19.04              | 17.03              | 16.96              | 16.86                |
| Baseline Controls               | NO                 | YES                | YES                | YES                | YES                  |
| Additional Controls             | NO                 | NO                 | YES                | YES                | YES                  |

Dependent Variable is Log (1 + Births Yu et al.). Standard errors (clustered at the NUTS 2 region level) in parentheses. Period Dummies and City FE always included.

Table A.13: Commune and Births: Distinguishing Positive and Negative Transitions

|                            | (1)<br>Positive<br>OLS | (2)<br>2SLS       | (3)<br>Negative<br>OLS | (4)<br>2SLS      |
|----------------------------|------------------------|-------------------|------------------------|------------------|
| Commune                    | 0.049***<br>(0.017)    | 0.171*<br>(0.098) | 0.066***<br>(0.022)    | 0.118<br>(0.117) |
| Fstat, instrum., 1st stage |                        | 13.56             |                        | 8.57             |
| Observations               | 2,812                  | 2,717             | 2,665                  | 2,569            |
| Adjusted R-squared         | 0.380                  | 0.203             | 0.363                  | 0.173            |

In this Table we estimate the effect of institutional transitions in the two directions separately. When estimating the effect of entry into Commune, we drop the city-century observations following a negative transition (from Commune=1 back to Commune=0). When studying the effect of exits, we drop the city-year observations prior to a positive transition (from Commune=0 to Commune=1). Dependent Variable is Log (1 + Births). Standard errors (clustered at the NUTS 2 region level) in parentheses. Period Dummies, City FE and full set of city-level controls always included. \*p<0.1, \*\* p<0.05, \*\*\* p<0.01.

Table A.14: Summary Statistics for the Dependent Variable in the Gravity Model

| Variable                 | Mean  | (Std. Dev.) | Min. | Max. | Percent of Zeros | N      |
|--------------------------|-------|-------------|------|------|------------------|--------|
| Immigrants (dyadic obs.) | 0.008 | (0.13)      | 0    | 14   | 99.42            | 366763 |