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**The Effect of Education on Mortality and Health: Evidence from a Schooling Expansion in Romania**

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**The Effect of Education on Mortality and Health:  
Evidence from a Schooling Expansion in Romania**

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**Abstract\***

This paper examines a schooling expansion in Romania that increased educational attainment for successive cohorts born between 1945 and 1950. We use a regression discontinuity design at the day level based on school entry cutoff dates to estimate impacts on mortality with 1994-2016 Vital Statistics data and self-reported health with 2011 Census data. We find that the schooling reform led to significant increases in years of schooling, higher employment rates, and reductions in fertility, but did not affect mortality, hospitalizations, or self-reported health. These estimates provide new evidence on the causal relationship between education and mortality outside of high-income countries and at lower margins of educational attainment.

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## 1. Introduction

There is substantial evidence showing that more educated people have better health and longer life expectancies. However, whether this correlation reflects a causal relationship remains an open question. A number of recent papers have used changes in compulsory schooling requirements to identify the causal impact of schooling on mortality in the United States (Lleras-Muney, 2005; Mazumder, 2008), the United Kingdom (Clark and Royer, 2013; Davies, et al. 2016), France (Albouy and Lequien, 2009), the Netherlands (van Kippersluis, et al., 2011), Sweden (Meghir, et al., 2018), and Taiwan (Kan, 2016). While this empirical approach can be compelling, the findings have been mixed and sometimes contradictory, even when based on the same educational expansions. Moreover, all of these studies are focused on high income countries where compulsory schooling laws usually affect students enrolled in secondary school. As a result, we know relatively little about the causal effect of education on health and mortality in low or middle-income countries, and at lower margins of educational attainment.

This paper examines the impact of a schooling expansion in Romania during the late 1950s and early 1960s, which sought to provide all students with at least 7 years of compulsory education. We show that successive cohorts of individuals, born between 1945 and 1950, who were affected by this schooling expansion, experienced rising educational attainment. Then we use a regression discontinuity design at the day level to compare individuals born just before the school entry cutoff of January 1 to those born just after, who were almost identical in age but began school later and therefore had greater opportunities to extend their education. Since students born immediately before and after January 1 were also the oldest and youngest in their respective classes,

we also draw on cohorts born after the schooling expansion had concluded to separate the effect of increased education from that of relative age and starting school younger.<sup>1</sup>

We demonstrate that the schooling expansion led to significant increases in years of schooling for the affected cohorts born between 1945 and 1950. This increase in educational attainment was accompanied by significant increases in labor force participation and decreases in fertility for women. Nevertheless, using detailed information on deaths from Vital Statistics data between 1994 and 2016, we do not find evidence that the schooling expansion reduced the mortality of affected cohorts up to the age of 71. Nor are there reductions in mortality from more specific causes of death. We also examine two health outcomes that may affect quality of life, as well as life expectancy: the total number of days spent in hospital (overall and by specific cause of hospitalization) based on Romanian in-patient registers from 1997 to 2017, and a measure of self-reported health problems using data from the 2011 Romanian Census. For both these outcomes, the estimated effects are small and insignificant, suggesting that the schooling reform had no discernable impacts on health.

Our findings indicate that more education does not help individuals avoid or postpone deaths during middle and old age. This is consistent with the null results in the most recent papers by Clark and Royer (2013) and Meghir et al. (2018) for the United Kingdom and Sweden. However, to the best of our knowledge, this is the first paper to provide compelling estimates for the causal relationship between education and mortality outside of high-income countries and at lower margins of educational attainment. We do not interpret these estimates as an argument against further

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<sup>1</sup> See Cascio and Schanzenbach (2016) for evidence on the impacts of relative age in Tennessee and Black, Devereux, and Salvanes (2011) for evidence on the effect of starting school younger in Norway.

educational expansions in the developing world. But they do suggest the need to be more circumspect about the potential for such expansions to improve health and increase life expectancy, at least at lower margins of educational attainment.

The paper is organized as follows. Section 2 reviews the related literature. Section 3 provides a background of the Romanian educational system and the educational expansion. Section 4 describes the data and the empirical strategy. Section 5 presents the main results, and Section 6 concludes.

## **2. Related Literature**

This section reviews some of the previous literature estimating the causal impact of education on health and mortality. We begin with a discussion of studies that take advantage of changes in compulsory schooling requirements. Then we describe some of the alternative empirical approaches used for identifying the causal effect of education at higher margins of educational attainment. For more detailed reviews of these and other studies, see Grossman (2006), Mazumder (2012) and Galama et al. (2018).

For the United States, Lleras-Muney (2005) uses Census data to examine the impact of changes in compulsory schooling laws between 1915 and 1939 that affected students over 14 years of age. Her instrumental variables (IV) estimates indicate that an additional year of schooling leads to significant declines in the probability of dying in the next 10 years. In a follow-up study, Mazumder (2008) notes that these results are not robust to including state-specific trends but presents evidence from the Survey of Income and Program Participation (SIPP) showing positive impacts of education on self-reported health status. Relatedly, Black et al. (2016) argue that virtually all of the variation in mortality rates is captured by cohort effects and state effects, making it

difficult to reliably estimate the effects of changing educational attainment due to state-level changes in compulsory schooling.<sup>2</sup>

For the United Kingdom, Clark and Royer (2013) use changes to British compulsory schooling laws in 1947 and 1972 that increased the minimum school leaving age from 14 to 15 and then from 15 to 16. Their regression discontinuity (RD) design does not provide strong evidence for an impact of education on mortality or other health outcomes. Davies et al. (2016) re-examine the 1972 change in compulsory schooling using UK Biobank data and find a statistically significant decline in mortality but their results are somewhat sensitive to functional form.

Other studies are mostly focused on European countries: For Sweden, Meghir et al. (2018) do not find improvements in mortality and other health measures for affected cohorts following an educational reform in Sweden that raised the number of years of compulsory schooling from 7/8 to 9, eliminated early selection based on academic ability, and introduced a national curriculum. Arendt (2005) and Albouy and Lequien (2009) also find no statistically significant impact of compulsory school reforms on health outcomes in Denmark or mortality in France, respectively. Yet van Kippersluis et al. (2011) do find that increasing compulsory school beyond grade 6 in the Netherlands leads to significant reduction in mortality in old age. For Taiwan, which looks like a developed country by most measures, Kan (2016) find that the extension of compulsory education from 6 to 9 years reduced men's mortality rate but did not affect women's mortality. In addition, a paper contemporaneous with ours by Dursun et al. (2018) examines the effect of a Turkish schooling expansion on health, but not on mortality,

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<sup>2</sup> In a paper that considers the effect of school quality on health, Aaronson et al. (2017) find that childhood exposure to Rosenwald schools in the Jim Crow south increased life expectancy, after accounting for the negative effects of migration.

A different set of studies use draft avoidance behavior in the United States during the Vietnam War to estimate the impact of college education on mortality and health outcomes. Buckles et al. (2016) show that the increased college going among men in cohorts associated with greater draft avoidance also leads to lower mortality in subsequent years. Grimard and Parent (2007) and de Walque (2007) use a similar identification strategy to estimate impacts on smoking behavior and find evidence suggesting that more education reduces the take-up of smoking and current smoking. The fact that the causal impact of education on mortality at the margin of a college education appears to differ from the impact at the margin of compulsory schooling suggests that looking at another margin of educational attainment could be informative as well.

In our own review of the literature, and in those by Grossman (2006), Mazumder (2012) and Galama et al. (2018), we have not found any papers that provide compelling causal estimates for the impact of education on mortality in low and middle-income countries and at lower margins of schooling.<sup>3</sup>

### **3. Background on Education in Romania**

During the post-war period, the structure and the organization of education in Romania was largely based on the model in the Soviet Union as codified by Decree No. 175 of 1948 (Braham, 1972).<sup>4</sup> There were several different types of schools. First, there were 4-year primary schools that offered grades 1 through 4 and were often located in rural areas. Second, there were 7-year general schools, called gymnasiums, which

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<sup>3</sup> Note, we have limited our review of the literature to the effects of education on own health and mortality. A separate literature has explored the impact of parental education on similar child outcomes; e.g. McCrary and Royer (2011) and Chou et al. (2010).

<sup>4</sup> This section relies heavily on information provided in Barham (1963, 1972).

offered grades 1 through 7 (and later expanded to grade 8), with the first four years covering similar material as in the 4- year primary schools. Third, there were 11-year schools, which offered grades 1 through 11 in one school. In accordance with the Romanian convention of referring to the first four grades as primary education, we consider graduates of gymnasiums as students who have completed *some* secondary education.

After a successful campaign to provide basic literacy education targeted towards all ages in the late 1940s and early 1950s, the government focused its attention on increasing enrollment beyond the first four grades. According to Giurescu et al. (1971, p. 351), the five year plan of 1955-1960 specified that the extension of compulsory schooling to 7 years was to be given special attention by the party and government. Thus, the directives of the Communist Party's Second Congress of 1955 which outlined the second five year plan, envisioned a "situation under which, by 1960-1961, the fifth grade would enroll 90 percent of the 4-year school graduates, and under which, according to the Third Five Year plan, the 7-year school would be universal and compulsory. At first only the first four grades were made compulsory, but villages and rural communities having 7-year schools were required by virtue of Decision No. 1035/1958 to make the 7 year schooling period universal beginning with the 1958-1959 academic year" (Braham, 1963).

Nevertheless, this process was not immediate and was constrained by a lack of enough schools offering 7 years of compulsory schooling: "Since this governmental action applied only to places where 7-year schools already existed, it appears that the extension of free compulsory education is to a large extent only nominal. Furthermore, with rural communities retaining the 4-year compulsory level, the lack of detailed planning to elevate their schools to the 7-year compulsory level has left an irregular



pattern of schooling in the provinces” (Braham, 1963). Filipescu and Oprea (1972) also confirm the gradual process of expanding education at the gymnasium level. They explain that the expansion of 7-year compulsory education began in 1956 within towns and larger villages that already had schools beyond the 4th grade, and that it gradually expanded until it was close to universal by 1961-1962.

We can document some of these changes using aggregate administrative data on enrollment from the Annual Statistics of the Socialist Republic of Romania. Figure 1 shows the large increase in the number of students graduating from gymnasium between 1955 and 1965. During this period, graduation from gymnasiums increased sharply from 116,698 in 1959 to 329,739 in 1963 and stayed at similar levels through the late 1960s and early 1970s.

Further evidence for these dramatic changes can be observed at the cohort level. By law, students entered grade 1 in September of the year following the calendar year in which they reached 6 years of age. Thus, the cohort born in 1945 was 6 years of age in 1951, entered first grade in the fall of 1952, entered fifth grade in the fall of 1956 and would have graduated with 7 years of schooling in the spring of 1959. This cohort should be the first cohort that could have been affected by the policy reform. Similarly, the cohort born in 1947 was the first cohort to have potentially benefited from the 1958 Government Decision that made 7-year of schooling compulsory. Finally, the cohort that entered fifth grade in 1961-1962, which according to Filipescu and Oprea (1972) is the first cohort to have achieved universal 7-year compulsory education, was born in 1950.

Figure 2 shows the highest educational attainment by year of birth for cohorts of individuals in the Romanian Census of 1992. There is a sharp decline in the proportion of individuals with primary education between cohorts born in 1945 and 1950. At the same time, we observe a sharp increase in the proportion of individuals who have

secondary education (which includes graduates of gymnasiums). Note that cohorts born between 1935 and 1944 also experienced large increases in educational attainment. This is mainly driven by the early literacy and education campaigns introduced after the Communist government came to power.

In Figure 3 we plot the “residual” percent of individuals born between 1943 and 1955 who completed primary education by their month of birth, after accounting for calendar month of birth effects. A number of interesting patterns emerge from this graph. First, and consistent with the results in Figure 2, we observe the large decrease in the proportion of students who have only primary education for those born between 1945 and 1950. Secondly, and more importantly for our empirical strategy, the decreases in percent of students with only primary education occur discontinuously, with disproportionately large decreases for those born after January 1st in this period. The discontinuities are especially visible for those born around January 1st of 1945, 1947, 1948 and 1949 and to a smaller extent for those born around January 1st of 1946 and 1950. At the same time, no similar discontinuities are visible for the control cohorts born between 1951 and 1953. The patterns in Figure 3 suggest that we can use detailed information on date of birth to estimate the impact of these educational expansions using a regression discontinuity design.

To summarize, the evidence on graduation rates from gymnasiums in the aggregate data coincides with the cohort analysis of educational attainment in the 1992 Census; and both are broadly consistent with the historical record of educational reforms in Romania. Together, they indicate that education levels past the first 4 years of primary schooling started to expand in the 1956-1957 school-year and by 1961-1962, enrollment in the 5th grade was essentially universal. In other words, the expansion affected cohorts born starting in 1945 and universal gymnasium education

was essentially completed for cohorts born after 1950. While we do not have information about the quality of education (and peer quality) from this period, we will examine the impact of this schooling reform on other important outcomes such as employment and fertility.

## **4. Data and Empirical Strategy**

### **4.1 Data**

Our main sample consists of individuals born in Romania between 1945 and 1953. Those born from 1945-1950 were enrolled in the affected grades during the period of schooling expansion while those born from 1951-1953 were enrolled after the expansions had already been completed. We use the three subsequent cohorts born immediately after the end of the schooling expansion to pick up the independent effect of relative age in our RD estimates.<sup>5</sup> We decided not to use the cohorts born prior to 1945 because they were directly affected by World War II, although our main results are similar if we use them instead. We put together information on these cohorts from several different datasets.

We use the 1992 Romanian Census, when individuals were 40 to 48 years of age, to estimate the impact of the schooling reform on educational attainment, certain labor market outcomes, and conduct specification checks of our empirical strategy.<sup>6</sup> Two features make this dataset especially useful for our analysis: First, with 35,000 to 45,000 observations in each yearly birth cohort, we have sufficient power to employ a

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<sup>5</sup> The three subsequent cohorts born immediately after the end of the schooling expansion are most similar in age to the cohorts affected by the schooling expansion and offer sufficiently large samples. However, our results are essentially unchanged when we use four, five, or six subsequent cohorts as our comparison group.

<sup>6</sup> This is a 15% random sample taken from the full Romanian Census by the Population Activities Unit (PAU) of the United Nations Economic Commission for Europe (UNECE).

regression discontinuity design. Second, there is detailed information about the day, month, and year of birth so we can identify the discontinuity induced by the policy within a narrow window.

The 1992 Census provides detailed information about the highest level of educational attainment for each respondent according to the following categories: none, primary, gymnasium, completed secondary education, post-secondary, and university education. For simplicity, we impute years of schooling by assigning the number of years associated with each level of education.<sup>7</sup> This serves as our main summary measure of education when estimating the impact of the schooling expansion. The Census also has information on socio-economic characteristics of our respondents, such as gender, ethnicity, and region of birth. We use these variables to validate our research design. Lastly, it contains information on labor force participation and occupational status (for those employed) as well as the fertility of women, which serve as useful auxiliary outcomes.

Panel A of Table 1 presents summary statistics for the individuals in cohorts born between 1945 and 1953. The average age at the time of the 1992 census is 42.2 years and the fraction of female respondents is almost exactly half. Almost 90 percent of the sample is ethnic Romanian, with about 7 percent ethnic Hungarians, and about 1.5 percent are Roma. The average imputed years of schooling in our sample is 9.58 years.

We use the 1994-2016 Vital Statistics Mortality files (VSM) to estimate the impact of the schooling expansion on mortality. These individual-level data cover the universe of deceased persons in Romania with detailed information on the day of birth/death and

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<sup>7</sup> We also use data collected by the Romanian National Statistics Institute in 1995 and 1996 with reports of actual years of schooling (rather than educational attainment) in order to validate our imputed measure of years of schooling. These data come from surveys based on the 1994 World Bank's Living Standards Measurement Studies (LSMS) for Romania.

the main cause of death, as well as some socio-economic characteristics.<sup>8</sup> Thus, we can observe mortality for the cohorts used in our analysis between the ages of 42 and 71 by day and year of birth.<sup>9</sup> We compute mortality by day of birth as follows: (i) we sum the number of deaths at each day of birth from 1945 to 1953 over the period 1994-2016; (ii) we estimate the population at risk by calculating the number of people alive in 1992 at each day of birth from 1945 to 1953;<sup>10</sup> then we take the ratio of (i) to (ii). This yields a mortality rate by day of birth which is at the finest level of our running variable.

Our calculation of the mortality rate could differ from the true mortality because of migration in and out of Romania. However, the number of immigrants (for the cohorts we study here) is close to zero and should not affect our results. Moreover, the VSM files include all people deceased abroad as long as they still have a Romanian residence and/or citizenship. Therefore, our mortality files should account for the majority of the Romanian migrants abroad who are temporary emigrants and do not change their permanent residence.<sup>11</sup> Still, we will directly examine the potential for bias due to migration by checking whether schooling expansion affects the probability of migration.

The VSM file provides detailed information on the main cause of death (ICD codes) so we are able to look separately at deaths associated with circulatory diseases and cancer. These are the two most important causes of death in Romania, accounting for 44.6% and 26.5% respectively of all deaths. Similar to Meghir et al. (2018) we also classify diseases according to the epidemiological literature as preventable and treatable;

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<sup>8</sup> The information on day of birth and death is from official records (death certificates, identity cards).

<sup>9</sup> Lleras-Muney (2005) and Clark and Royer (2013) suggest that the largest effects of education on mortality occur before the age of 64. Life expectancy in Romania was 69.5 years in 1994, 74.2 in 2011, and 75.5 years in 2016.

<sup>10</sup> We use sample weights to calculate the total population because we only have a 15% census sample.

<sup>11</sup> According to Statistics Romania these emigrants are the vast majority (over the 95%) of emigrants.

preventable causes of death may reflect health behaviors while the treatable causes of death may be related to access to healthcare.<sup>12</sup>

Panel B of Table 1 shows the overall mortality rate and the mortality rate by category for our main sample. Approximately 26 percent of our sample died between 1994-2016. The largest category of deaths was due to circulatory diseases which account for 10.5 percentage points, followed by cancer at 7.7 percentage points; preventable deaths accounted for 5.9 percentage points, while treatable diseases only for 3.9 percentage points.

We use the 1997-2017 National Inpatient Registers to calculate the number of nights spent in hospital care by day of birth. The National Inpatient register contains individual-level data on duration and ICD codes for all hospital stays in Romanian hospitals starting in January 1, 1997. Based on 7,892,000 hospital entries for our cohorts of interest, we calculate that individuals in our cohorts aged 54 to 72 spent an average of 25.6 days in hospital, as shown in Panel C of Table 1.

Finally, the 2011 Romanian Census all respondents are asked whether they have any health related problems that may affect their daily life at work, school, at home, etc. Thus, we can compute a measure of self-reported health for individuals who survived until 2011. Approximately 7.6 percent of people in our cohorts of interest reported having such problems. Those who answered affirmatively were given a set of six follow-up questions – whether they were (i) visually, (ii) hearing, or (iii) movement impaired, (iv) whether they had any memory or concentration problems, (v) self-care or (vi) difficulties in communication with their peers.

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<sup>12</sup> We use the ICD 10 codes for defining cancer, circulatory diseases and treatable and preventable causes of death. See the Notes at the end of the tables for more information.

## 4.2 Empirical Strategy

As described earlier, the schooling expansions in Romania occurred over a five-year period from 1956 to 1961 and affected those born between 1945 and 1950. Since the government rapidly expanded access to schooling during this period, a child born just after January 1 would have benefited from the additional schools slots created by the government over the course of a year, as compared to a child born just before January 1 who would have been part of an earlier cohort. Indeed, the discontinuities in the fraction of individuals whose highest level of education was primary school were clearly visible in Figure 3 for the years 1945, 1947, 1948 and 1949. In this section, we estimate these discontinuities more formally using a regression discontinuity (RD) design.

We estimate the differences across successive cohorts during the period of educational expansion (i.e. in the “treatment years” of 1945-1950) using the following equation:

$$y_i = \beta' X_i + \alpha AFTER_i + f(day_i) + \varepsilon_i \quad (1)$$

where  $y_i$  is an outcome such as education, mortality or health for individual  $i$ ,  $X_i$  is a set of control variables,  $AFTER_i$  is an indicator for individuals born just after the school entry cutoff of January 1, and  $f(day_i)$  is a parametric or non-parametric function of the day of birth which serves as our running variable. For simplicity, our preferred specifications do not include any control variables except for a constant, although including them does not affect our results. The coefficient on  $\alpha$  is an estimate for the effect of being born just after the school entry cutoff on the relevant outcome. When the outcome is a measure of education, such as years of schooling, it represents a “first

stage” estimate; when the outcome is a measure of mortality or health, it represents the “reduced-form” estimate.

If we assume that the exclusion restriction holds (i.e. that being born after the school entry cutoff affects mortality only through years of schooling), the ratio of the reduced-form and first stage coefficients represents an estimate for the impact of education on mortality. However, the exclusion restriction may not hold since those individuals born just after the school entry cutoff are generally the oldest children in their class; that is, if relative age has an independent effect on health or mortality.

In order to account for any independent effect of relative age, we also compare individuals who were born just before and after the school entry cutoff in a period without educational expansion (i.e. in the “control years” of 1951-1953). We do this by estimating a regression equation similar to equation (1) above using this set of control years. But we also estimate regression models that directly compare the impact of being born just after the school entry cutoff in treatment years to control years:

$$y_i = \beta' X_i + \alpha AFTER_i + \gamma TREAT_i + \delta AFTER_i * TREAT_i + f(day_i) + \varepsilon_i \quad (2)$$

where  $TREAT_i$  is an indicator for individuals born during years of educational expansion 1945-1950, and the other variables are defined as before (with some abuse of notation). In this specification, the coefficient on the interaction term,  $\delta$ , yields the impact of being born just after the school entry cutoff during treatment years over and above the effect in control years that did not experience educational expansions, assuming that the effect of relative age does not vary over time.<sup>13</sup>

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<sup>13</sup> This specification is similar to ones used by other recent papers which estimate a difference in RD discontinuities across cohorts. Grembi et al. (2016) provide a more formal presentation of the standard assumptions underlying this setting.



A key consideration when implementing a regression discontinuity design is the functional form of the forcing variable,  $f(day_i)$ . We present estimates using a local linear regression as suggested by Hahn, Todd, and van der Klaauw (2001). The choice of the window is somewhat arbitrary as we need to strike a balance between the advantages of having more precise estimates with larger windows and mitigating the possibility of confounding time effects with more narrow windows. Therefore, for our main tables we present specifications using a 180, 120, 90, 60 and 30 day intervals, as well as the Calonico, Cattaneo and Titiunik (CCT) optimal bandwidth which is a refinement of the Imbens and Kalyanarman (2012) optimal bandwidth (Calonico, Cattaneo and Titiunik, 2014). We also confirm that our results are robust to using parametric specifications that include higher order polynomials such as linear, quadratic and cubic trends in day of birth (results available by request). All regressions cluster on day of birth in order to avoid the problems associated with specification error in the case of discrete covariates (Lee and Card, 2008).

A common specification check for the regression discontinuity design is to verify that the density of observations is continuous around the cutoff (McCrary, 2008). When we examine the density, we find substantial heaping on January 1 and on some of the days immediately preceding it.<sup>14</sup> We believe that this heaping is mainly due to delays in the reporting of births that occurred during the holiday period between Christmas and

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<sup>14</sup> These density tests are shown in Appendix Table 1 and Appendix Figure 1. They are structured in a similar fashion to the main tables as described in the results section.

New Year's Day when government offices were closed.<sup>15</sup> There is also some heaping that occurs on the first day of every month, perhaps due to reporting errors.<sup>16</sup>

Insofar as this type of heaping is similar for our “treatment” and “control” years, we can account for this issue in the regression that uses both sets of years. However, we also attempt to deal with this issue using a “donut-RD” design as suggested by Barreca, et al. (2016). In particular, we present all of our results when dropping individuals born within 7 days of January 1 in order to be symmetric around the cutoff. Results are qualitatively similar when we exclude individuals born more than one week before or after January 1 or when we exclude individuals born only one or several days before January 1 (available by request).<sup>17</sup>

## **5. Results**

### **5.1 Effects on educational attainment**

We begin by estimating the impact of the schooling expansion on years of completed schooling based on the level of education recorded in the 1992 Census. These “first stage” results are shown in Table 2 which has three panels: Panel A presents estimates for  $\alpha$  from equation (1) using the treatment years, 1945-1950; Panel B presents estimates for  $\alpha$  from equation (1) using the control years, 1951-1953; Panel C presents estimates for  $\alpha$  and  $\delta$  from our preferred specification (2) which includes both treatment and control years. Columns (1) to (6) in each panel show estimates for

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<sup>15</sup> In contrast to most other orthodox denominations, Christmas always remained on December 25 for the Romanian Orthodox. Consistent with this explanation, it appears the spike in observations occurs on January 2 in years when January 1 is a Sunday.

<sup>16</sup> Barreca et al. (2016) also document some heaping at the beginning of each month in the California Vital Statistics records used by McCrary and Royer (2011).

<sup>17</sup> We also verify that our available covariates vary smoothly around the discontinuity in Appendix Tables 2 and 3. With a few exceptions, the coefficients are small and insignificant.

alternative bandwidths. These include 180, 120, 90, 60 and 30 days of the January 1 cutoff, as well as the optimal bandwidth proposed in Calonico, Cattaneo and Titiunik (2014). Columns (7) to (12) show analogous specifications that exclude observations within 7 days of the January 1 cutoff (i.e. 7-day donut-RD regressions).

Panel A of Table 2 indicates that each successive cohort during the school expansion period 1945-1950 received an additional 1/5 to 3/5 years of schooling; the point estimates for the impact of being born just after vs. just before the January 1 cutoff in the treatment years range from 0.21 to 0.67 years of schooling using our different bandwidths. In contrast, the estimates in Panel B showing the impact of being born just after vs. just before January 1 in the control years of 1951-1953 are small and statistically insignificant in all specifications. Panel C shows estimates from the specification that combines both treatment and control years. In these specifications, the impact of the school expansion is captured by  $AFTER_i * TREAT_i$  and shows impacts of 0.23 to 0.57 years of schooling, all highly significant. The results using the donut specifications are about 30% smaller in magnitude but still statistically significant in all specifications.<sup>18</sup> The range of these estimate is not altogether surprising given the large number of different specifications that we consider. However, we take our preferred specification to be the CCT bandwidth for the full sample, implying a first stage effect of approximately a 1/2 year of schooling.

We also present our “first stage” results graphically in Figure 4. Panels A, C and E plot average years of schooling by day of birth for individuals born six months before and after January 1st of each year; panels B, D and F plot the same data by week of birth,

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<sup>18</sup> Appendix Table 4 uses the 1994-1996 LSMS datasets to estimate the impact of the schooling expansion on reported years of schooling rather than an imputed measure based on completed educational levels. The results are somewhat less precise but generally similar to those in Table 2.

which often makes it easier to discern the patterns. The graphs are normalized so that day 1 corresponds to January 1 and week 1 corresponds to the week of January 1 to January 7, and the fitted lines are based on linear spline regressions.

Panels A and B show a clear discontinuity after January 1 for the treatment years of 1945-1950. This visual evidence confirms that individuals born merely a couple of days apart received a substantially different amount of schooling as a result of the school expansion. In contrast, panels C and D of Figure 4 reveal no change in average educational attainment before and after January 1<sup>st</sup> in the control cohort. Nevertheless, each of the first four panels in Figure 4 show some time trends, consistent with the presence of seasonality in the timing of births. Such time effects are not visible in Panels E and F of Figure 4, which use both treatment and control years to estimate a version of equation (2) that differences out the impacts in the control years from those in the treatment years.

## **5.2 Do the effects on education matter? Impacts on employment and fertility**

Despite the clear impacts of the schooling reform on educational attainment, one might question whether the increase in the quantity of education actually mattered for later outcomes. For example, though the Romanian government attempted to keep the quality of education high, it is possible that it declined during a period rapid expansion of education. We are not able to check for changes in the quality of education directly. However, before examining the potential impact of our school expansion on health and mortality, we examine whether the expansion had an impact on other relevant outcomes, such as labor market or fertility related outcomes.

Table 3 presents estimates for the impact of the schooling expansions on labor force participation, measured as an indicator for being employed at the time of the 1992

Census.<sup>19</sup> This table is structured similarly to the other tables, with Panels A, B, and C showing impacts for the treatment years, control years, and all the years together. The impact of the schooling expansion on employment is most clearly visible with the full sample in Panel A, which shows that individuals entering school in successive cohorts are 1.0 to 3.6 percent points more likely to be employed.<sup>20</sup> Given that the average employment rate for this sample is 0.83, these estimates correspond to an increase of 1.2 to 4.3 percent. The impacts on employment are in a narrower range of 1.1 to 1.7 percentage points in Panel C that combines both treatment and control years, although we lose precision in the narrowest bandwidths. The patterns for the donut regressions are broadly consistent despite further losses in precision. A graphical depiction of these impacts can be seen in Panels A, B, E and F of Appendix Figure 2.

We also observe significant impacts of Romania's schooling expansion on women's fertility, as shown in Table 4. Our preferred estimates reported in Panel C, using the full sample of women, show that exposure to the expansion led to decreased fertility in the range of 0.08 to 0.29 children. Again, the estimates in the donut regressions are less precise but remain marginally significant in all the specifications except for the largest bandwidth. A graphical depiction of these impacts can be seen in Panels A, B, E and F of Appendix Figure 3. It is also worth noting that the impacts of the schooling reform on years of schooling and employment remain similar when restricting the sample to women.

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<sup>19</sup> Note that, although registered unemployment was essentially zero under the Communist regime through 1990, it rose to over 10% by 1992 (Earle and Pauna, 1996). Unfortunately, the 1992 Census does not contain any information about earnings or income.

<sup>20</sup> Moreover, we also found some impacts on occupational outcomes, such as the likelihood of working in a manual occupation or the skill level associated with one's occupation. These results are available by request.

To conclude, the estimated impacts of the educational expansion on labor market and fertility outcomes confirm that the schooling expansion had consequential effects on a range of socio-economic outcomes, easing concerns about the quality of the education.

### **5.3 Effects on mortality**

This section examines whether the school expansion policy had an impact on mortality, our main outcome of interest in this paper. We focus on the mortality rate calculated from Vital Statistics data between 1994 and 2016, as described earlier. Table 5, which has the same structure as the previous tables, reveals no evidence of a statistically significant effect of being born just after vs. just before the January 1 cutoff on mortality for the treatment years of 1945-1950 (in Panel A) or for the control years of 1951-1953 (in Panel B), except for the smallest bandwidths. Furthermore, all of the significant effects disappear once we consider the donut regression that excludes individuals born 7 days before and after January 1.

We see a couple of marginally significant coefficients on  $AFTER_i * TREAT_i$  in Panel C that includes both treatment and control years, although these have positive signs. Still, with 10 out of the 12 point estimates from our preferred specification in Panel C not showing any statistically significant effect, we conclude that there is no evidence for an impact of the schooling expansion on mortality. Given the standard errors for our interacted specifications using the full sample, we can rule out with 95% confidence that the schooling expansions reduced mortality by more than 1.7 percentage points between 1994-2016 when the average mortality rate was 26 percent.

A graphical analysis of the mortality results is presented in Figure 5, structured similarly to the preceding figures. The patterns in Panels A-F provide a visual

interpretation of the regression estimates from Table 5. We do not see evidence for large discontinuities in the mortality rate between 1994 and 2016 and, if anything, they point against the finding that education reduces mortality.

We also consider the effect of the schooling expansion on specific causes of death. We first focus on mortality from the two most common causes of death in Romania: cancer and circulatory diseases. The regression estimates for these causes of death are shown in Tables 6 and 7 respectively, while the figures are shown in Appendix Figure 4 and 5 respectively. We also classify certain causes of death as preventable or treatable, similar to Meghir et al. (2018). The regression estimates for these causes of death are shown in Appendix Table 5 and 6 respectively. In none of the tables do we observe evidence for a consistent effect of the schooling expansion on mortality. Similarly, none of the corresponding graphs show visible discontinuities around the regression discontinuity cutoffs. Thus, we do not find any more evidence for the impact of the schooling expansion on specific causes of death than on the mortality rate as a whole.

#### **5.4 Impacts on hospitalizations and self-reported health problems**

In addition to the impact of the schooling expansion on mortality, we examine its effect on less extreme measures of health that may affect quality of life: hospitalizations based on in-patient registers from 1997 to 2017, and self-reported health problems taken from the 2011 Romanian Census. We focus on the total number of days spent in hospital and present these results in Table 8 and Appendix Figure 6, which are structured in a similar fashion to the previous tables and figures. Overall, we do not find

an impact of the schooling expansion on the number of days spent in hospital.<sup>21</sup> Nor are there significant impacts on the number of hospitalizations, or on the duration of hospitalizations by specific cause such as cancer and circulatory diseases (results available on request).

The impacts on self-reported health problems are shown in Table 9 and Appendix Figure 7. Again, we do not find an impact of the schooling expansion on self-reported health problems among individuals who survived until 2011, although there are a few marginally significant coefficients in Panel A using the full sample. It is worth noting that, since we have access to the full 100% sample of the 2011 Census, these results are estimated with substantial precision. Given our standard errors for our interacted specifications using the full sample, we can rule out with 95% confidence that the schooling expansions reduced the fraction of people with health related problems by more than 0.6 percentage points, which corresponds to 0.023 standard deviation units. We have also explored the specific dimensions of health problems associated with our variable of self-reported health (i.e. vision, hearing, impaired movement, memory, self-care and communication) and did not find any meaningful impacts for these specific categories.<sup>22</sup>

## 5.5 Robustness

We consider a number of robustness checks of our main results. These include the following: (i) we only consider cohorts born 1949-1952 that are relatively similar in age; (ii) we drop the 1945 cohort due to the possibility of being affected in utero by

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<sup>21</sup> There are significant effects when using the narrowest 30 day bandwidth for the full sample but these appear to be outliers and disappear in the donut regressions.

<sup>22</sup> We do not observe any significant differences in the estimated impacts by gender or by birth in urban vs. rural regions for any of our mortality and health outcomes.



WWII; (iii) we separately examine mortality between 1994-2005 and 2006-2016; and (iv) we look at mortality rates for people aged 52-62 only. Our findings remain qualitatively unchanged in each of these alternative specifications. These results are available upon request.

To address concerns about bias due to migration, we consider whether our school expansion directly affected the probability of external migration. The 2011 census contains information on all persons who migrated abroad for a period of at least 12 months (at the time of the census). Hence, the vast majority of the Romanian emigrants are covered i.e., all individuals working abroad who maintain their houses, identity cards or/and remain registered by the Romanian administrative bodies.<sup>23</sup> Using a similar strategy as before, we show in Appendix Table 7 that the effects are similar for both the treatment and control years. Thus, there is no overall impact of the schooling expansion on the likelihood of the individuals (who survived until 2011) to have emigrated.

The migration results presented above, while reassuring, are not able to capture any possible effects of the schooling expansion on permanent migration. We address this possibility through an indirect test. Using information from the 1992 and 2011 census samples, we calculate the (weighted) number of people born in a given day who are in the 2011 census as a (weighted) fraction of the number in the 1992 census. This ratio should capture a combination of both mortality and migration between 1992-

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<sup>23</sup> According to Statistics Romania, about 95% of the Romanian emigrants are temporary migrants, meaning that they keep their Romanian ID's. Moreover, the death of these individuals is reported in the Romanian Mortality Files. While permanent migrants who do not remain registered by the Romanian administrative bodies are not covered, we believe this is a second-order issue because these are mostly highly educated migrants (university or more) who, most likely, were not affected by our policy.

2011. These results are presented in Appendix Table 8, and confirm that there is no impact of the school expansion on this combined measure of mortality and migration.

## 5.6 Mechanisms

Our findings indicate that the Romanian schooling expansion did not improve health or reduce mortality, while it did have some positive impacts on labor market and fertility related outcomes. In this section, we attempt to explore some of the mechanisms underlying these findings. However, insofar as education can impact health and mortality through many different channels, our discussion remains largely speculative.

First, more education may lead to higher income and perhaps better health care. While Romania has universal access to the public healthcare system independently of the individual income, financial resources may still be important because of informal payments (i.e. bribes). Our main results did suggest that the schooling expansion led to greater labor market opportunities (e.g. higher employment) but the Census data did not include information on income. In Appendix Table 9, we use the LSMS survey to examine whether the impact of the school expansion affected income and found positive but insignificant effects. Note that, using the LSMS data, we find positive and significant impacts of the schooling expansion on employment, similar to the results using Census data.<sup>24</sup>

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<sup>24</sup> Education could also affect mortality through changes in the occupation structure. Indeed, we observe some evidence that Romania's schooling expansion, shifted individuals out of manual jobs and farming and into technicians and professional jobs. However, whether these changes should have led to improved health is not completely clear if more education enables individuals to find work in more skilled occupations, with better working conditions, we might expect to find positive health impacts. However, some skilled occupations may be associated with more stress than certain less skilled occupations. Moreover, it is possible that some relatively skilled manufacturing jobs may have worse working conditions than jobs in the informal sector such as agriculture.

Second, even if more education would lead to higher incomes, the impact of income on health is not obvious. Income could allow individuals to access better health care, but it may also lead to an increased consumption of unhealthy goods, such as alcohol and cigarettes. This seems to be the case in Romania where, using the Romanian Household Budget Survey, we find positive and significant correlations between education and smoking. However, when we attempted to estimate our regression discontinuity specifications using this data, we find no significant effects of education on smoking behavior (see Appendix Table 10).<sup>25</sup> Using the same data, we also find no effects on the likelihood of having a chronic condition.<sup>26</sup>

Thus, our analysis does not yield any strong conclusions about the role of particular mechanisms in explaining our results.<sup>27</sup> However, these results need to be interpreted with care since they are mostly based on imprecise estimates using small ancillary datasets.

## **6. Conclusion**

This paper analyzes a schooling expansion in Romania, which aimed to ensure that all students received at least 7 years of compulsory schooling. The schooling expansion affected five consecutive cohorts born between 1945-1950 and we use a regression discontinuity (RD) design to estimate impacts by comparing the differences across successive cohorts of affected students. We find that beginning school in a (one year)

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<sup>25</sup> Specifically, we use the 2001-2009 Romanian Household Budget Survey (RHBS) which is a national representative survey, covering about 30,000 households each year and contains detailed socio-economic information on all household members. Note that the RHBS data does not have the day of birth, but only the month and year and therefore we cannot show the donuts specifications.

<sup>26</sup> The RHBS data also showed no effect on the likelihood of being hospitalized or on the number of days hospitalized during the last 30 days (results available by request).

<sup>27</sup> Given the findings in Aaronson et al. (2017), we also examined the role of internal migration. However, we did not find significant effects of the schooling expansion on internal migration, measured as an indicator for whether the person lives in the locality of birth in 2011 (results available upon request).

later cohort increases educational attainment by approximately a 1/2 year of schooling. We do not find any consistent significant impacts of the schooling reform on mortality, hospitalizations, or self-reported health. Moreover, we can rule out that the schooling expansions reduced mortality by more than 1.7 percentage points between 1994 and 2016 or that they reduced self-reported health problems by more than 0.02 standard deviation units for the full sample of individuals in the affected cohorts.

Whether education causally affects health and mortality is an important question for both developed and developing countries alike. However, most of the previous work has focused on the United States and Western Europe. The findings in this literature are mixed and there is not strong evidence that education significantly improves health or decreases mortality. We extend the literature by estimating causal impacts for a population that is substantially poorer and also experienced changes at a lower margin of educational attainment. Our findings indicate the absence of a causal effect of education health and mortality, even in this setting. While we have attempted to examine the underlying mechanisms for these findings, more work needs to be done to better understand why we do not observe a strong relationship between education and health across a variety of different settings.

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Table 1: Summary Statistics

	Mean	S.D.	Obs
<b>Panel A: Census data</b>			
Female	0.506	0.500	375,925
Age	42.225	2.575	375,925
<i>Ethnicity</i>	.	.	-
Romanian	0.893	0.309	375,911
Hugarian	0.074	0.261	375,911
Roma	0.015	0.122	375,911
Other	0.018	0.135	375,911
Years of schooling	9.580	3.661	374,778
Employed	0.830	0.376	375,925
Number of children	2.564	1.798	190,248
Self-reported health index (2011)	0.076	0.265	2,058,787
<b>Panel B: Mortality data</b>			
Overall mortality	0.260	0.081	3,284
<i>Mortality by category</i>			
Cancer	0.077	0.026	3,284
Circulatory	0.105	0.039	3,284
Preventable	0.059	0.018	3,284
Treatable	0.039	0.014	3,284
<b>Panel C: Hospitalization data</b>			
Time hospitalized (days)	25.884	6.973	3,284

Source: 1992 Romanian Census (PAU sample), 2011 Romanian Census, In-patient registry data & Romania VSM files



Table 2: 1st Stage Effects of Educational Expansion on Years of Schooling

<i>dependent variable: years of schooling</i>		Full sample						Excluding 7 days on each side of cutoff					
		bandwidth (days)	180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)
<b>Panel A: Treated years</b>													
After	0.210*** [0.047]	0.300*** [0.058]	0.397*** [0.067]	0.520*** [0.082]	0.675*** [0.124]	0.644*** [0.114]	0.106*** [0.039]	0.162*** [0.049]	0.245*** [0.060]	0.340*** [0.079]	0.333** [0.134]	0.371*** [0.098]	
Sample size	234,035	150,855	109,086	69,652	32,700	38,694	225,296	142,116	100,347	60,913	23,961	40,947	
R-squared	0.020	0.019	0.019	0.018	0.018	0.017	0.020	0.019	0.019	0.018	0.016	0.017	
<b>Panel B: Control years</b>													
After	-0.023 [0.049]	-0.003 [0.066]	0.033 [0.080]	0.069 [0.106]	0.101 [0.165]	0.064 [0.101]	-0.056 [0.045]	-0.045 [0.061]	-0.008 [0.076]	0.020 [0.102]	-0.075 [0.176]	0.016 [0.109]	
Sample size	135,608	89,656	65,889	42,433	19,981	45,524	130,612	84,660	60,893	37,437	14,985	33,706	
R-squared	0.001	0.001	0.001	0.000	0.000	0.000	0.001	0.001	0.000	0.000	0.000	0.000	
<b>Panel C: All years</b>													
After	-0.023 [0.049]	-0.003 [0.066]	0.033 [0.080]	0.069 [0.106]	0.101 [0.165]	0.079 [0.119]	-0.056 [0.045]	-0.045 [0.061]	-0.008 [0.076]	0.020 [0.102]	-0.075 [0.176]	0.013 [0.121]	
After*Treatment	0.233*** [0.057]	0.304*** [0.073]	0.364*** [0.086]	0.451*** [0.103]	0.573*** [0.127]	0.483*** [0.110]	0.162*** [0.058]	0.207*** [0.078]	0.252** [0.097]	0.320** [0.130]	0.408* [0.214]	0.356*** [0.154]	
Sample size	369,643	240,511	174,975	112,085	52,681	94,070	355,908	226,776	161,240	98,350	38,946	74,277	
R-squared	0.022	0.020	0.020	0.019	0.018	0.019	0.022	0.021	0.020	0.019	0.018	0.018	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calomico, Cattaneo, and Triunlik (2014) optimal bandwidth.

Table 3: Effect of Educational Expansion on Employment

		<i>dependent variable: nonemployment</i>											
		Full sample						Excluding 7 days on each side of cutoff					
bandwidth (days)		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.010*** [0.004]	0.018*** [0.004]	0.021*** [0.005]	0.027*** [0.006]	0.037*** [0.007]	0.031*** [0.007]	0.003 [0.003]	0.009** [0.004]	0.011** [0.005]	0.015** [0.007]	0.024 [0.015]	0.016* [0.009]	
Sample size	234,737	151,310	109,428	69,865	32,803	48,622	225,968	142,541	100,659	61,096	24,034	47,304	
R-squared	0.002	0.002	0.002	0.002	0.003	0.002	0.002	0.002	0.002	0.002	0.002	0.002	
<b>Panel B: Control years</b>													
After	-0.005 [0.005]	0.001 [0.006]	0.006 [0.006]	0.011 [0.008]	0.023** [0.010]	0.016* [0.008]	-0.010* [0.005]	-0.005 [0.007]	0.001 [0.008]	0.007 [0.012]	0.038* [0.020]	0.010 [0.013]	
Sample size	136,022	89,943	66,106	42,569	20,045	34,204	131,004	84,925	61,088	37,551	15,027	31,470	
R-squared	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.000	
<b>Panel C: All years</b>													
After	-0.005 [0.005]	0.001 [0.006]	0.006 [0.006]	0.011 [0.008]	0.023** [0.010]	0.022** [0.009]	-0.010* [0.005]	-0.005 [0.007]	0.001 [0.008]	0.007 [0.012]	0.038* [0.020]	0.010 [0.013]	
After*Treatment	0.015*** [0.005]	0.017** [0.007]	0.015** [0.008]	0.016* [0.009]	0.015 [0.012]	0.011 [0.011]	0.013** [0.006]	0.014* [0.008]	0.010 [0.010]	0.008 [0.014]	-0.015 [0.026]	0.006 [0.016]	
Sample size	370,759	241,253	175,534	112,434	52,848	68,436	356,972	227,466	161,747	98,647	39,061	82,459	
R-squared	0.002	0.002	0.002	0.002	0.003	0.002	0.002	0.002	0.002	0.002	0.002	0.002	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Table 4: Effect of Educational Expansion on Fertility

<i>dependent variable: number of children</i>		Full sample						Excluding 7 days on each side of cutoff						
		bandwidth (days)	180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)	CCT (12)
<b>Panel A: Treated years</b>														
After	-0.033 [0.023]	-0.060** [0.027]	-0.088*** [0.030]	-0.117*** [0.033]	-0.116*** [0.032]	-0.115*** [0.033]	-0.012 [0.027]	-0.036 [0.037]	-0.070 [0.047]	-0.113* [0.067]	-0.140 [0.138]	-0.108 [0.089]		
Sample size	119,711	76,884	55,805	35,818	16,772	37,194	115,410	72,583	51,504	31,517	12,471	21,924		
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.000	0.000	0.000	0.001	0.001		
<b>Panel B: Control years</b>														
After	0.046 [0.036]	0.082* [0.046]	0.096* [0.054]	0.124* [0.066]	0.169* [0.089]	0.168** [0.078]	0.024 [0.039]	0.059 [0.054]	0.067 [0.070]	0.091 [0.100]	0.239 [0.213]	0.128 [0.120]		
Sample size	67,945	44,902	32,951	21,130	9,907	14,323	65,514	42,471	30,520	18,699	7,476	14,991		
R-squared	0.001	0.001	0.001	0.001	0.002	0.002	0.001	0.001	0.001	0.001	0.002	0.002		
<b>Panel C: All years</b>														
After	0.046 [0.036]	0.082* [0.046]	0.096* [0.054]	0.124* [0.066]	0.169* [0.089]	0.127* [0.067]	0.024 [0.039]	0.059 [0.054]	0.067 [0.070]	0.091 [0.100]	0.239 [0.213]	0.078 [0.090]		
After*Treatment	-0.079** [0.038]	-0.142*** [0.047]	-0.184*** [0.055]	-0.241*** [0.065]	-0.285*** [0.087]	-0.245*** [0.066]	-0.036 [0.042]	-0.095* [0.057]	-0.137* [0.073]	-0.203* [0.105]	-0.379* [0.225]	-0.180* [0.094]		
Sample size	187,656	121,786	88,756	56,948	26,679	54,891	180,924	115,054	82,024	50,216	19,947	57,495		
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.002	0.001		

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Table 5: Effects of Educational Expansion on Mortality Rate

		Full sample						Excluding 7 days on each side of cutoff					
<i>dependent variable: mortality rate</i>		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.010 [0.009]	0.006 [0.011]	0.003 [0.013]	0.003 [0.015]	-0.006 [0.016]	0.002 [0.014]	0.011 [0.010]	0.007 [0.014]	0.003 [0.018]	0.004 [0.026]	-0.016 [0.053]	0.008 [0.030]	
Sample size	2,154	1,434	1,074	714	354	900	2,070	1,350	990	630	270	540	
R-squared	0.118	0.112	0.105	0.108	0.130	0.105	0.124	0.119	0.113	0.118	0.147	0.119	
<b>Panel B: Control years</b>													
After	-0.011 [0.007]	-0.012 [0.008]	-0.015 [0.009]	-0.020* [0.011]	-0.046*** [0.015]	-0.020* [0.011]	-0.002 [0.007]	0.004 [0.009]	0.006 [0.010]	0.013 [0.012]	-0.001 [0.022]	0.011 [0.011]	
Sample size	1,077	717	537	357	177	348	1,035	675	495	315	135	369	
R-squared	0.071	0.073	0.079	0.082	0.146	0.083	0.058	0.055	0.055	0.046	0.042	0.048	
<b>Panel C: All years</b>													
After	-0.011 [0.007]	-0.012 [0.008]	-0.015 [0.009]	-0.020* [0.011]	-0.046*** [0.015]	-0.017* [0.010]	-0.002 [0.007]	0.004 [0.009]	0.006 [0.010]	0.013 [0.012]	-0.001 [0.022]	0.013 [0.012]	
After*Treatment	0.021* [0.012]	0.018 [0.014]	0.018 [0.016]	0.023 [0.019]	0.040* [0.020]	0.019 [0.018]	0.013 [0.014]	0.003 [0.018]	-0.003 [0.023]	-0.009 [0.032]	-0.014 [0.062]	-0.006 [0.035]	
Sample size	3,231	2,151	1,611	1,071	531	1,260	3,105	2,025	1,485	945	405	855	
R-squared	0.253	0.242	0.232	0.219	0.223	0.223	0.261	0.252	0.242	0.229	0.228	0.224	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calomico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Table 6: Effects of Educational Expansion on Mortality Rate due to Cancer

		<i>dependent variable: mortality rate</i>											
		Full sample						Excluding 7 days on each side of cutoff					
bandwidth (days)		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After		0.001 [0.003]	0.001 [0.004]	-0.000 [0.005]	-0.001 [0.005]	-0.003 [0.005]	-0.000 [0.005]	0.001 [0.004]	-0.000 [0.005]	-0.001 [0.007]	-0.002 [0.010]	-0.012 [0.019]	-0.003 [0.012]
Sample size		2,154	1,434	1,074	714	354	900	2,070	1,350	990	630	270	498
R-squared		0.062	0.058	0.055	0.063	0.098	0.058	0.067	0.063	0.060	0.066	0.094	0.069
<b>Panel B: Control years</b>													
After		-0.003 [0.002]	-0.002 [0.003]	-0.003 [0.004]	-0.004 [0.004]	-0.014** [0.006]	-0.004 [0.004]	0.001 [0.003]	0.003 [0.003]	0.004 [0.004]	0.008 [0.005]	-0.000 [0.009]	0.007 [0.005]
Sample size		1,077	717	537	357	177	420	1,035	675	495	315	135	339
R-squared		0.049	0.047	0.049	0.049	0.104	0.048	0.039	0.036	0.035	0.028	0.022	0.029
<b>Panel C: All years</b>													
After		-0.003 [0.002]	-0.002 [0.003]	-0.003 [0.004]	-0.004 [0.004]	-0.014** [0.006]	-0.004 [0.004]	0.001 [0.003]	0.003 [0.003]	0.004 [0.004]	0.008 [0.005]	-0.000 [0.009]	0.007 [0.005]
After**Treatment		0.004 [0.004]	0.003 [0.006]	0.003 [0.006]	0.004 [0.008]	0.011 [0.008]	0.003 [0.007]	0.000 [0.005]	-0.003 [0.007]	-0.006 [0.009]	-0.010 [0.013]	-0.012 [0.024]	-0.009 [0.014]
Sample size		3,231	2,151	1,611	1,071	531	1,287	3,105	2,025	1,485	945	405	801
R-squared		0.167	0.160	0.153	0.145	0.163	0.147	0.174	0.169	0.162	0.152	0.158	0.148

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. We use the ICD-10 diseases codes - chapter C for cancer. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Table 7: Effects of Educational Expansion on Mortality Rate due to Circulatory Diseases

		Full sample						Excluding 7 days on each side of cutoff					
<i>dependent variable: mortality rate</i>		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.007** [0.004]	0.005 [0.004]	0.003 [0.005]	0.002 [0.006]	-0.003 [0.007]	0.002 [0.006]	0.009** [0.004]	0.007 [0.006]	0.005 [0.007]	0.006 [0.010]	0.005 [0.019]	0.009 [0.011]	
Sample size	2,154	1,434	1,074	714	354	756	2,070	1,350	990	630	270	528	
R-squared	0.223	0.216	0.210	0.215	0.238	0.214	0.227	0.220	0.215	0.222	0.255	0.223	
<b>Panel B: Control years</b>													
After	-0.003 [0.002]	-0.003 [0.003]	-0.004 [0.003]	-0.005 [0.003]	-0.012*** [0.004]	-0.006 [0.003]	-0.000 [0.003]	0.000 [0.004]	0.001 [0.004]	0.003 [0.005]	-0.004 [0.010]	0.002 [0.005]	
Sample size	1,077	717	537	357	177	345	1,035	675	495	315	135	378	
R-squared	0.123	0.128	0.137	0.144	0.201	0.145	0.108	0.105	0.108	0.100	0.095	0.104	
<b>Panel C: All years</b>													
After	-0.003 [0.002]	-0.003 [0.003]	-0.004 [0.003]	-0.005 [0.003]	-0.012*** [0.004]	-0.005 [0.003]	-0.000 [0.003]	0.000 [0.004]	0.001 [0.004]	0.003 [0.005]	-0.004 [0.010]	0.002 [0.006]	
After* Treatment	0.010** [0.004]	0.008 [0.005]	0.007 [0.006]	0.007 [0.007]	0.009 [0.007]	0.007 [0.006]	0.010* [0.005]	0.006 [0.007]	0.004 [0.008]	0.004 [0.011]	0.009 [0.020]	0.006 [0.012]	
Sample size	3,231	2,151	1,611	1,071	531	1,197	3,105	2,025	1,485	945	405	819	
R-squared	0.410	0.399	0.391	0.383	0.388	0.385	0.413	0.402	0.393	0.382	0.380	0.376	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth. We use the ICD-10 diseases codes- chapter I for the circulatory diseases.

Table 8: Effects of Educational Expansion on Hospitalization Days

		<i>dependent variable: number of days in hospital</i>											
		Full sample						Excluding 7 days on each side of cutoff					
<i>bandwidth (days)</i>		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.613 [0.784]	0.183 [0.962]	0.129 [1.105]	0.031 [1.280]	-0.900 [1.402]	0.066 [1.229]	0.725 [0.915]	0.209 [1.238]	0.163 [1.575]	0.086 [2.228]	-2.717 [4.379]	0.119 [2.705]	
Sample size	2,154	1,434	1,074	714	354	816	2,070	1,350	990	630	270	492	
R-squared	0.010	0.014	0.016	0.019	0.050	0.019	0.009	0.012	0.012	0.014	0.053	0.019	
<b>Panel B: Control years</b>													
After	-0.779 [0.848]	-1.336 [1.029]	-1.837 [1.158]	-2.403* [1.398]	-6.368*** [1.824]	-2.252* [1.322]	0.535 [0.855]	0.614 [1.049]	0.759 [1.185]	1.892 [1.384]	-1.651 [2.452]	1.759 [1.551]	
Sample size	1,077	717	537	357	177	402	1,035	675	495	315	135	255	
R-squared	0.008	0.014	0.022	0.022	0.096	0.021	0.007	0.007	0.009	0.006	0.013	0.006	
<b>Panel C: All years</b>													
After	-0.779 [0.848]	-1.336 [1.029]	-1.837 [1.158]	-2.403* [1.397]	-6.368*** [1.822]	-2.262* [1.330]	0.535 [0.855]	0.614 [1.049]	0.759 [1.184]	1.892 [1.383]	-1.651 [2.448]	1.721 [1.355]	
After* Treatment	1.392 [1.200]	1.519 [1.465]	1.966 [1.664]	2.433 [1.968]	5.467** [2.375]	2.304 [1.891]	0.191 [1.300]	-0.405 [1.688]	-0.597 [2.067]	-1.805 [2.762]	-1.066 [5.493]	-1.711 [2.666]	
Sample size	3,231	2,151	1,611	1,071	531	1,188	3,105	2,025	1,485	945	405	999	
R-squared	0.013	0.017	0.020	0.020	0.064	0.020	0.014	0.017	0.017	0.014	0.045	0.014	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level, respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Table 9: Effects of Educational Expansion on Self-reported Health Problems

		Full sample						Excluding 7 days on each side of cutoff					
<i>dependent variable: any self-reported problem</i>		180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)	CCT (12)
<b>Panel A: Treated years</b>													
After	-0.002* [0.001]	-0.003* [0.002]	-0.004* [0.002]	-0.004* [0.002]	-0.003 [0.002]	-0.003 [0.002]	-0.003 [0.002]	-0.001 [0.001]	-0.002 [0.002]	-0.002 [0.002]	-0.002 [0.002]	-0.002 [0.004]	-0.002 [0.002]
Sample size	1,247,177	809,041	586,100	370,899	171,223	294,219	1,202,917	764,813	541,893	326,718	127,065	495,227	
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001
<b>Panel B: Control years</b>													
After	0.000 [0.001]	-0.002 [0.001]	-0.003* [0.002]	-0.003 [0.002]	-0.003 [0.003]	-0.004* [0.002]	0.001 [0.001]	-0.001 [0.002]	-0.003 [0.002]	-0.005* [0.003]	-0.008 [0.005]	-0.004 [0.002]	
Sample size	777,061	515,496	379,811	242,960	113,734	221,478	749,024	487,483	351,807	214,965	85,749	240,759	
R-squared	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
<b>Panel C: All years</b>													
After	0.000 [0.001]	-0.002 [0.001]	-0.003* [0.002]	-0.003 [0.002]	-0.003 [0.003]	-0.004 [0.002]	0.001 [0.001]	-0.001 [0.002]	-0.003 [0.002]	-0.005* [0.003]	-0.008 [0.005]	-0.003 [0.002]	
After* <sup>Treatment</sup>	-0.002 [0.002]	-0.001 [0.002]	-0.001 [0.002]	-0.000 [0.003]	0.000 [0.003]	0.001 [0.003]	-0.002 [0.002]	-0.001 [0.003]	0.001 [0.003]	0.003 [0.004]	0.006 [0.008]	0.001 [0.003]	
Sample size	2,024,238	1,324,537	965,911	613,859	284,957	470,322	1,951,941	1,252,296	893,700	541,683	212,814	793,613	
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, \* and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.



Appendix Table 1: Density checks

		Full sample						Excluding 7 days on each side of cutoff					
		180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)	CCT (12)
<b>Panel A: Treated years</b>													
After	0.112*** [0.027]	0.140*** [0.038]	0.156*** [0.050]	0.196*** [0.071]	0.330*** [0.121]	0.481*** [0.167]	0.066*** [0.010]	0.073*** [0.011]	0.062*** [0.011]	0.054*** [0.011]	0.077*** [0.020]	0.081*** [0.019]	
Sample size	2,160	1,440	1,080	720	360	204	2,076	1,356	996	636	276	312	
R-squared	0.204	0.230	0.229	0.252	0.394	0.541	0.253	0.328	0.346	0.368	0.473	0.423	
<b>Panel B: Control years</b>													
After	0.102*** [0.021]	0.124*** [0.030]	0.135*** [0.039]	0.160*** [0.056]	0.258** [0.098]	0.399*** [0.143]	0.071*** [0.009]	0.078*** [0.010]	0.071*** [0.011]	0.056*** [0.013]	0.065** [0.027]	0.050** [0.020]	
Sample size	1,080	720	540	360	180	96	1,038	678	498	318	138	210	
R-squared	0.215	0.268	0.271	0.277	0.377	0.531	0.218	0.340	0.381	0.423	0.448	0.412	
<b>Panel C: All years</b>													
After	0.102*** [0.021]	0.124*** [0.030]	0.135*** [0.039]	0.160*** [0.056]	0.258** [0.097]	0.394*** [0.142]	0.071*** [0.009]	0.078*** [0.010]	0.071*** [0.011]	0.056*** [0.013]	0.065** [0.027]	0.067** [0.030]	
After*Treatment	0.010 [0.008]	0.016 [0.011]	0.021 [0.013]	0.037** [0.017]	0.071*** [0.026]	0.106*** [0.032]	-0.005 [0.005]	-0.006 [0.007]	-0.009 [0.008]	-0.002 [0.010]	0.012 [0.018]	0.006 [0.019]	
Sample size	3,240	2,160	1,620	1,080	540	306	3,114	2,034	1,494	954	414	378	
R-squared	0.208	0.242	0.242	0.259	0.390	0.550	0.247	0.338	0.365	0.396	0.483	0.492	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Appendix Table 2: Specification Tests for Covariates (1)

bandwidth (days)	Full sample						Excluding 7 days on each side of cutoff					
	180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)	CCT (12)
<b>Panel A: Female</b>												
After	-0.025*** [0.010]	-0.038*** [0.012]	-0.050*** [0.014]	-0.073*** [0.017]	-0.104*** [0.023]	-0.102*** [0.022]	-0.009 [0.007]	-0.016 [0.010]	-0.024** [0.012]	-0.047*** [0.016]	-0.100*** [0.025]	-0.082*** [0.022]
After*Treatment	-0.013* [0.008]	-0.022** [0.010]	-0.028** [0.011]	-0.024* [0.014]	-0.027 [0.020]	-0.027 [0.019]	-0.006 [0.009]	-0.011 [0.011]	-0.016 [0.014]	-0.001 [0.019]	0.049 [0.031]	0.030 [0.026]
Sample size	370,759	241,253	175,534	112,434	52,848	54,726	356,972	227,466	161,747	98,647	39,061	50,723
R-squared	0.001	0.001	0.002	0.003	0.006	0.006	0.000	0.001	0.001	0.001	0.002	0.002
<b>Panel B: Ethnic Romanian</b>												
After	0.015*** [0.005]	0.024*** [0.006]	0.030*** [0.006]	0.033*** [0.008]	0.033*** [0.011]	0.033*** [0.009]	0.013** [0.005]	0.023*** [0.007]	0.029*** [0.008]	0.031*** [0.011]	0.021 [0.022]	0.030*** [0.012]
After*Treatment	-0.008 [0.005]	-0.011 [0.007]	-0.013 [0.008]	-0.013 [0.010]	-0.008 [0.014]	-0.011 [0.011]	-0.008 [0.005]	-0.013* [0.007]	-0.017** [0.009]	-0.020* [0.012]	-0.005 [0.023]	-0.018 [0.013]
Sample size	370,746	241,245	175,527	112,430	52,845	92,262	356,959	227,458	161,740	98,643	39,058	88,604
R-squared	0.001	0.001	0.001	0.001	0.002	0.002	0.001	0.001	0.001	0.001	0.001	0.002
<b>Panel C: Ethnic Hungarian</b>												
After	-0.013*** [0.003]	-0.018*** [0.004]	-0.021*** [0.004]	-0.023*** [0.005]	-0.021*** [0.007]	-0.023*** [0.005]	-0.013*** [0.004]	-0.019*** [0.005]	-0.024*** [0.006]	-0.027*** [0.008]	-0.018 [0.014]	-0.026*** [0.008]
After*Treatment	0.005 [0.004]	0.008 [0.006]	0.010 [0.007]	0.013 [0.008]	0.011 [0.012]	0.013 [0.008]	0.005 [0.005]	0.010 [0.006]	0.013 [0.008]	0.018* [0.011]	0.005 [0.020]	0.018 [0.011]
Sample size	370,746	241,245	175,527	112,430	52,845	116,860	356,959	227,458	161,740	98,643	39,058	88,604
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.002	0.002	0.002

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Appendix Table 3: Specification Tests for Covariates (2)

bandwidth (days)	Full sample						Excluding 7 days on each side of cutoff					
	180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (7)	120 (8)	90 (9)	60 (10)	30 (11)	CCT (12)
<b>Panel A: Ethnic Roma</b>												
After	-0.001 [0.002]	-0.002 [0.002]	-0.004 [0.003]	-0.005 [0.004]	-0.007 [0.005]	-0.005 [0.004]	0.001 [0.002]	-0.001 [0.003]	-0.003 [0.003]	-0.002 [0.004]	0.006 [0.008]	-0.001 [0.005]
After* <i>Treatment</i>	0.001 [0.002]	0.001 [0.003]	0.001 [0.003]	-0.001 [0.004]	-0.004 [0.007]	-0.003 [0.005]	0.002 [0.002]	0.001 [0.003]	0.001 [0.003]	-0.002 [0.004]	-0.013* [0.007]	-0.003 [0.005]
Sample size	370,746	241,245	175,527	112,430	52,845	78,270	356,959	227,458	161,740	98,643	39,058	86,490
R-squared	0.000	0.001	0.001	0.001	0.001	0.001	0.000	0.000	0.000	0.000	0.001	0.000
<b>Panel B: Ethnic Other</b>												
After	-0.002 [0.002]	-0.004 [0.002]	-0.004* [0.003]	-0.004 [0.003]	-0.006 [0.004]	-0.004 [0.003]	-0.001 [0.002]	-0.003 [0.003]	-0.003 [0.003]	-0.002 [0.005]	-0.008 [0.010]	-0.003 [0.004]
After* <i>Treatment</i>	0.001 [0.002]	0.002 [0.003]	0.002 [0.003]	0.001 [0.004]	-0.000 [0.006]	0.001 [0.004]	0.002 [0.002]	0.003 [0.003]	0.004 [0.004]	0.004 [0.005]	0.013 [0.010]	0.004 [0.004]
Sample size	370,746	241,245	175,527	112,430	52,845	104,347	356,959	227,458	161,740	98,643	39,058	125,898
R-squared	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
<b>Panel C: Born in Bucharest</b>												
After	-0.009*** [0.003]	-0.011*** [0.003]	-0.012*** [0.004]	-0.011** [0.005]	-0.014** [0.007]	-0.012** [0.006]	-0.008*** [0.003]	-0.010*** [0.004]	-0.010** [0.004]	-0.006 [0.006]	-0.006 [0.012]	-0.010** [0.005]
After* <i>Treatment</i>	0.002 [0.003]	0.003 [0.004]	0.003 [0.005]	0.001 [0.006]	0.003 [0.009]	0.001 [0.007]	0.004 [0.003]	0.005 [0.004]	0.005 [0.005]	0.002 [0.008]	0.008 [0.016]	0.005 [0.006]
Sample size	370,759	241,253	175,534	112,434	52,848	82,189	356,972	227,466	161,747	98,647	39,061	142,984
R-squared	0.000	0.000	0.001	0.001	0.001	0.001	0.000	0.000	0.000	0.001	0.001	0.001

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. *Treatment* is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Appendix Table 4: Effects of Educational Expansion on Actual Years of Schooling

		<i>dependent variable: years of schooling</i>									
		1994 LSMS					1995-96 LSMS				
bandwidth (days)	6 (1)	5 (2)	4 (3)	3 (4)	2 (5)	6 (6)	5 (7)	4 (8)	3 (9)	2 (10)	
<b>Panel A: Treated years</b>											
After	0.291 [0.275]	0.550** [0.212]	0.641** [0.197]	0.981*** [0.149]	1.183*** [0.020]	0.175 [0.225]	0.356* [0.189]	0.495*** [0.126]	0.601*** [0.098]	0.837*** [0.014]	
Sample size	5,294	4,376	3,439	2,526	1,596	8,383	6,909	5,416	3,909	2,421	
R-squared	0.017	0.018	0.017	0.017	0.019	0.027	0.029	0.032	0.032	0.029	
<b>Panel B: Control years</b>											
After	0.107 [0.321]	0.455 [0.268]	0.471** [0.172]	0.565*** [0.111]	0.724*** [0.025]	-0.029 [0.194]	-0.110 [0.151]	-0.156 [0.250]	0.155** [0.044]	0.129*** [0.004]	
Sample size	2,356	1,968	1,550	1,143	690	4,887	4,068	3,209	2,377	1,498	
R-squared	0.001	0.001	0.004	0.007	0.006	0.004	0.006	0.003	0.006	0.003	
<b>Panel C: All years</b>											
After	0.107 [0.321]	0.455 [0.268]	0.471** [0.172]	0.565*** [0.111]	0.724*** [0.025]	-0.029 [0.194]	-0.110 [0.151]	-0.156 [0.250]	0.155** [0.044]	0.129*** [0.004]	
After* <sup>a</sup> Treatment	0.184 [0.267]	0.095 [0.270]	0.169 [0.197]	0.416 [0.210]	0.458*** [0.033]	0.204 [0.184]	0.466** [0.167]	0.651** [0.220]	0.446*** [0.102]	0.708*** [0.013]	
Sample size	7,650	6,344	4,989	3,669	2,286	13,270	10,977	8,625	6,286	3,919	
R-squared	0.040	0.042	0.046	0.050	0.044	0.041	0.043	0.046	0.043	0.042	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Appendix Table 5: Effects of Educational Expansion on Mortality Rate due to Preventable Diseases

		Full sample						Excluding 7 days on each side of cutoff					
<i>dependent variable: mortality rate</i>		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.000	-0.000	-0.001	-0.001	-0.001	-0.003	-0.001	0.001	0.000	-0.000	0.000	-0.003	-0.000
	[0.002]	[0.003]	[0.003]	[0.003]	[0.003]	[0.004]	[0.003]	[0.002]	[0.003]	[0.004]	[0.006]	[0.012]	[0.007]
Sample size	2,154	1,434	1,074	714	354	816	2,070	1,350	990	630	270	474	
R-squared	0.023	0.022	0.020	0.024	0.047	0.023	0.025	0.026	0.025	0.031	0.056	0.035	
<b>Panel B: Control years</b>													
After	-0.002	-0.002	-0.002	-0.003	-0.010*	-0.003	-0.000	0.002	0.003	0.005	0.002	0.005	
	[0.002]	[0.003]	[0.003]	[0.004]	[0.006]	[0.004]	[0.002]	[0.003]	[0.003]	[0.003]	[0.006]	[0.003]	
Sample size	1,077	717	537	357	177	390	1,035	675	495	315	135	330	
R-squared	0.025	0.025	0.026	0.027	0.067	0.026	0.019	0.021	0.021	0.017	0.016	0.017	
<b>Panel C: All years</b>													
After	-0.002	-0.002	-0.002	-0.003	-0.010*	-0.003	-0.000	0.002	0.003	0.005	0.002	0.004	
	[0.002]	[0.003]	[0.003]	[0.004]	[0.006]	[0.004]	[0.002]	[0.003]	[0.003]	[0.003]	[0.006]	[0.004]	
After*Treatment	0.002	0.001	0.001	0.002	0.007	0.002	0.001	-0.002	-0.004	-0.005	-0.005	-0.003	
	[0.003]	[0.004]	[0.004]	[0.005]	[0.006]	[0.005]	[0.003]	[0.005]	[0.006]	[0.008]	[0.014]	[0.009]	
Sample size	3,231	2,151	1,611	1,071	531	1,206	3,105	2,025	1,485	945	405	783	
R-squared	0.057	0.056	0.051	0.049	0.070	0.049	0.061	0.061	0.058	0.055	0.070	0.054	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth. The preventable causes of death include: Lung cancer (C33-C34), Cirrhosis of liver (K70, K74.3-K74.6), External causes of death (V, W, X, Y).

Appendix Table 6: Effects of Educational Expansion on Mortality Rate due to Treatable Diseases

		Full sample						Excluding 7 days on each side of cutoff					
<i>dependent variable: mortality rate</i>		180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
<b>Panel A: Treated years</b>													
After	0.002 [0.002]	0.001 [0.002]	-0.000 [0.002]	-0.001 [0.002]	-0.003 [0.003]	-0.001 [0.002]	0.002 [0.002]	0.002 [0.002]	0.001 [0.003]	0.001 [0.004]	-0.002 [0.008]	0.001 [0.004]	
Sample size	2,154	1,434	1,074	714	354	798	2,070	1,350	990	630	270	642	
R-squared	0.106	0.102	0.095	0.096	0.101	0.095	0.111	0.108	0.103	0.109	0.133	0.108	
<b>Panel B: Control years</b>													
After	-0.002 [0.001]	-0.002 [0.001]	-0.003** [0.002]	-0.004** [0.002]	-0.007*** [0.002]	-0.004** [0.002]	-0.000 [0.001]	-0.000 [0.002]	-0.001 [0.002]	0.001 [0.002]	0.001 [0.005]	0.000 [0.002]	
Sample size	1,077	717	537	357	177	363	1,035	675	495	315	135	336	
R-squared	0.045	0.045	0.051	0.055	0.094	0.054	0.038	0.031	0.031	0.027	0.026	0.027	
<b>Panel C: All years</b>													
After	-0.002 [0.001]	-0.002 [0.001]	-0.003** [0.002]	-0.004** [0.002]	-0.007*** [0.002]	-0.004** [0.002]	-0.000 [0.001]	-0.000 [0.002]	-0.001 [0.002]	0.001 [0.002]	0.001 [0.005]	0.000 [0.002]	
After* <sup>a</sup> Treatment	0.003 [0.002]	0.003 [0.002]	0.003 [0.003]	0.003 [0.003]	0.004 [0.004]	0.003 [0.003]	0.003 [0.002]	0.002 [0.003]	0.002 [0.004]	0.001 [0.005]	-0.003 [0.011]	0.001 [0.005]	
Sample size	3,231	2,151	1,611	1,071	531	1,098	3,105	2,025	1,485	945	405	1,017	
R-squared	0.224	0.218	0.211	0.200	0.191	0.201	0.228	0.224	0.216	0.208	0.201	0.209	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth. Treatable causes of death (cf. ICD 10) include: Tuberculosis (A15-A19, B90), Malignant neoplasm of cervix uteri (C53); Chronic rheumatic heart disease (I05-I09); All respiratory diseases (J00-J99); Asthma (J45, J46); Appendicitis (K35-K38); Abdominal hernia (K40-K46); Hypertensive and cerebrovascular disease (I10-I15, I60-I69); Cholelithiasis and cholecystitis (K80-K81).

Appendix Table 7: Effect of Educational Expansion on International Migration in the 2011 census

		Full sample						Excluding 7 days on each side of cutoff						
<i>dependent variable: migration rate</i>		bandwidth (days)	180	120	90	60	30	CCT	180	120	90	60	30	CCT
		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	
<b>Treated years</b>														
After	0.011***	0.010***	0.010***	0.012***	0.016***	0.011***	0.010***	0.008***	0.007**	0.007	0.003	0.007**	0.007**	0.003
	[0.002]	[0.002]	[0.003]	[0.004]	[0.006]	[0.003]	[0.002]	[0.003]	[0.003]	[0.005]	[0.008]	[0.003]	[0.003]	
Sample size	1,289,754	836,609	605,991	383,596	177,102	451,583	1,244,105	790,960	560,342	337,947	131,453	549,215		
R-squared	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	
<b>Control years</b>														
After	0.012***	0.014***	0.015***	0.018***	0.023***	0.018***	0.009***	0.011***	0.009**	0.009	0.007	0.009*	0.009*	
	[0.002]	[0.003]	[0.004]	[0.004]	[0.005]	[0.004]	[0.003]	[0.003]	[0.004]	[0.006]	[0.009]	[0.004]		
Sample size	806,943	535,308	394,535	252,495	118,210	263,465	777,836	506,201	365,428	223,388	89,103	332,173		
R-squared	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
<b>All years</b>														
After	0.012***	0.014***	0.015***	0.018***	0.023***	0.016***	0.009***	0.011***	0.009**	0.009	0.007	0.010***	0.010***	
	[0.002]	[0.003]	[0.004]	[0.004]	[0.005]	[0.004]	[0.003]	[0.003]	[0.004]	[0.006]	[0.009]	[0.004]		
After*Treatment	-0.001	-0.004	-0.005	-0.006	-0.007	-0.005	0.001	-0.003	-0.002	-0.003	-0.004	-0.003	-0.003	
	[0.003]	[0.004]	[0.004]	[0.005]	[0.007]	[0.005]	[0.003]	[0.004]	[0.005]	[0.007]	[0.012]	[0.005]		
Sample size	2,096,697	1,371,917	1,000,526	636,091	295,312	753,859	2,021,941	1,297,161	925,770	561,335	220,556	1,035,423		
R-squared	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	0.002	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.

Appendix Table 8: Effect of Educational Expansion on Attrition between the 1992 and 2011 census

		<i>dependent variable: attrition rate</i>											
		Full sample					Excluding 7 days on each side of cutoff						
bandwidth (days)		180 (1)	120 (2)	90 (3)	60 (4)	30 (5)	CCT (6)	180 (6)	120 (7)	90 (8)	60 (9)	30 (10)	CCT (6)
<b>Treated years</b>													
After	-0.008 [0.023]	-0.017 [0.027]	-0.020 [0.031]	-0.014 [0.036]	-0.024 [0.041]	-0.019 [0.035]	-0.011 [0.026]	-0.024 [0.034]	-0.032 [0.042]	-0.026 [0.059]	-0.071 [0.112]	-0.016 [0.063]	
Sample size	2,153	1,433	1,073	712	352	791	2,071	1,351	991	630	270	576	
R-squared	0.017	0.023	0.030	0.038	0.076	0.037	0.014	0.015	0.018	0.019	0.046	0.019	
<b>Control years</b>													
After	-0.059** [0.027]	-0.061* [0.032]	-0.065* [0.035]	-0.073* [0.039]	-0.161*** [0.047]	-0.077* [0.040]	-0.028 [0.030]	-0.012 [0.038]	0.004 [0.044]	0.046 [0.052]	0.002 [0.089]	0.036 [0.064]	
Sample size	1,077	717	537	356	176	335	1,036	676	496	315	135	222	
R-squared	0.015	0.023	0.028	0.023	0.063	0.023	0.009	0.017	0.021	0.016	0.011	0.013	
<b>All years</b>													
After	-0.059** [0.027]	-0.061* [0.032]	-0.065* [0.035]	-0.073* [0.039]	-0.161*** [0.047]	-0.073* [0.039]	-0.028 [0.030]	-0.012 [0.038]	0.004 [0.044]	0.046 [0.052]	0.002 [0.088]	0.028 [0.048]	
After* Treatment	0.050 [0.037]	0.044 [0.045]	0.045 [0.050]	0.059 [0.056]	0.137** [0.055]	0.060 [0.056]	0.018 [0.043]	-0.013 [0.056]	-0.036 [0.068]	-0.072 [0.091]	-0.074 [0.168]	-0.063 [0.081]	
Sample size	3,230	2,150	1,610	1,068	528	1,059	3,107	2,027	1,487	945	405	1,127	
R-squared	0.022	0.029	0.036	0.045	0.090	0.046	0.016	0.018	0.022	0.023	0.041	0.023	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion. The CCT bandwidth is the Calonico, Cattaneo, and Titiunik (2014) optimal bandwidth.



Appendix Table 9: Effects of Educational Expansion on Employment and Income in the 1995-2000 LSMS data

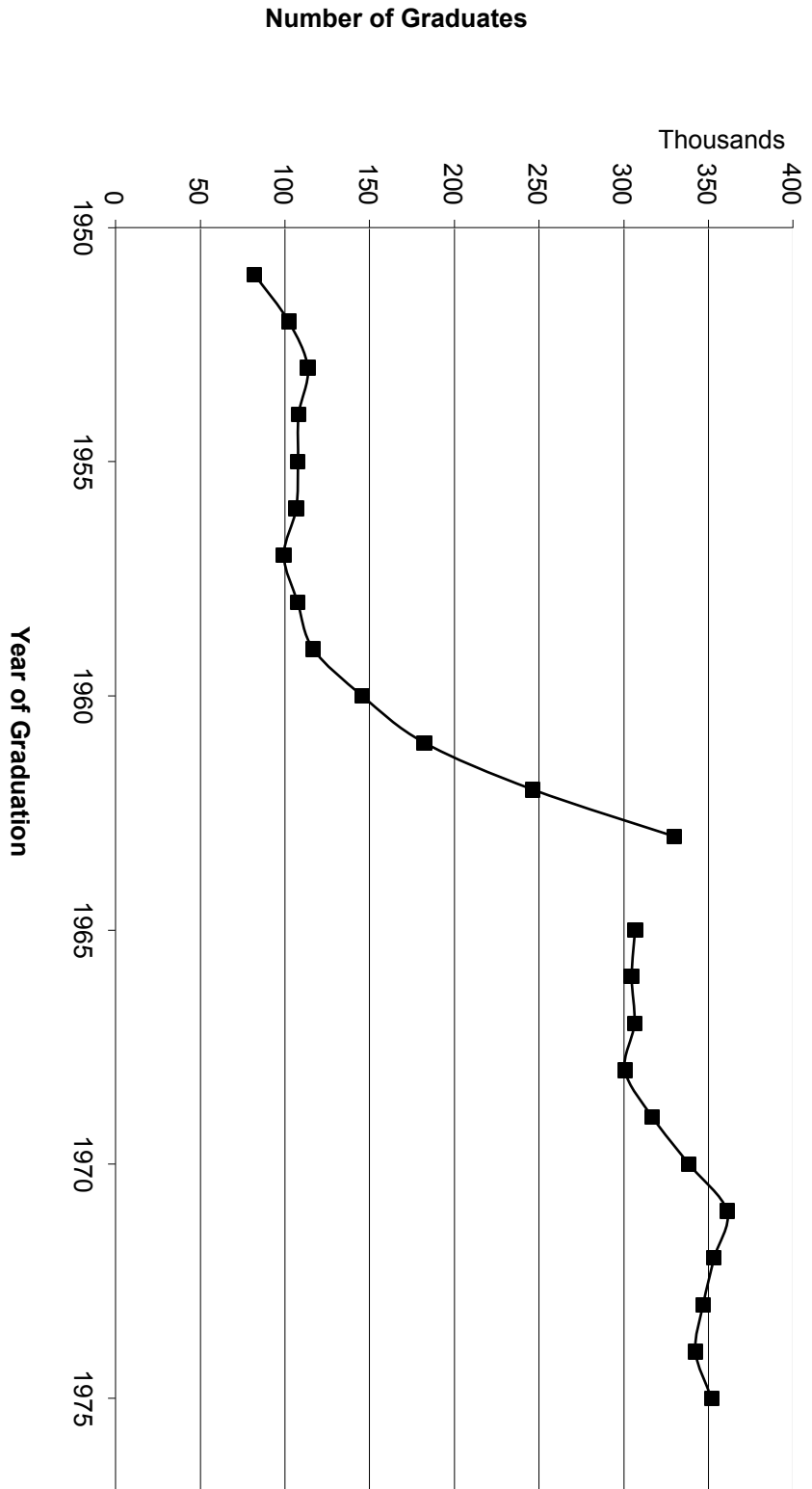
		Employment					Log Income				
bandwidth (days)		6 (1)	5 (2)	4 (3)	3 (4)	2 (5)	6 (6)	5 (7)	4 (8)	3 (9)	2 (10)
<b>Panel A: Treated years</b>											
After	-0.0006 [0.0120]	0.0127 [0.0096]	0.0188* [0.0094]	0.0227* [0.0096]	0.0341*** [0.0011]	-0.0130 [0.0260]	0.0079 [0.0232]	0.0122 [0.0266]	0.0299 [0.0209]	0.0675*** [0.0009]	
Sample size	40,558	33,512	26,211	19,089	12,005	40,234	33,245	26,012	18,955	11,920	
R-squared	0.016	0.016	0.015	0.014	0.013	0.010	0.010	0.010	0.009	0.010	
<b>Panel B: Control years</b>											
After	-0.0239* [0.0111]	-0.0132 [0.0112]	-0.0224 [0.0143]	-0.0188 [0.0130]	-0.0131*** [0.0015]	-0.0337 [0.0412]	-0.0275 [0.0426]	-0.0286 [0.0506]	0.0105 [0.0438]	0.0905*** [0.0021]	
Sample size	23,158	19,268	15,220	11,251	7,129	22,999	19,140	15,119	11,178	7,092	
R-squared	0.001	0.001	0.001	0.001	0.001	0.001	0.002	0.001	0.002	0.004	
<b>Panel C: All years</b>											
After	-0.0239* [0.0111]	-0.0132 [0.0112]	-0.0224 [0.0143]	-0.0188 [0.0130]	-0.0131*** [0.0015]	-0.0337 [0.0412]	-0.0275 [0.0426]	-0.0286 [0.0506]	0.0105 [0.0438]	0.0905*** [0.0021]	
After* <i>Treatment</i>	0.0234*** [0.0070]	0.0259*** [0.0082]	0.0412*** [0.0064]	0.0415*** [0.0044]	0.0473*** [0.0018]	0.0207 [0.0202]	0.0355 [0.0217]	0.0408 [0.0282]	0.0195 [0.0229]	-0.0230*** [0.0013]	
Sample size	63,716	52,780	41,431	30,340	19,134	63,233	52,385	41,131	30,133	19,012	
R-squared	0.021	0.021	0.020	0.020	0.019	0.010	0.010	0.010	0.009	0.011	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion.

Appendix Table 10: Effect of Educational Expansion on Smoking and Chronic Conditions

		Full sample - smoking					Full sample - chronic condition				
bandwidth (days)		(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
<b>Treated years</b>											
After	0.0092* [0.0051]	0.0092* [0.0044]	0.0095** [0.0040]	0.0117** [0.0041]	0.0067 [0.0045]	0.0099** [0.0044]	0.0101* [0.0050]	0.0107* [0.0054]	0.0116* [0.0054]	0.0151** [0.0050]	
Sample size		113,367	94,611	73,720	53,302	32,486	113,367	94,611	73,720	53,302	32,486
R-squared		0.003	0.003	0.003	0.003	0.003	0.001	0.001	0.001	0.001	0.001
<b>Control years</b>											
After	0.0017 [0.0060]	-0.0006 [0.0064]	0.0007 [0.0071]	0.0031 [0.0058]	-0.0004 [0.0030]	0.0017 [0.0063]	0.0016 [0.0060]	0.0040 [0.0051]	0.0032 [0.0047]	0.0087 [0.0054]	
Sample size		64,852	54,126	42,289	30,730	19,048	64,852	54,126	42,289	30,730	19,048
R-squared		0.001	0.001	0.001	0.001	0.001	0.001	0.000	0.001	0.001	
<b>All years</b>											
After	0.0017 [0.0060]	-0.0006 [0.0064]	0.0007 [0.0071]	0.0031 [0.0058]	-0.0004 [0.0030]	0.0017 [0.0063]	0.0016 [0.0060]	0.0040 [0.0051]	0.0032 [0.0047]	0.0087 [0.0054]	
After* <sup>†</sup> Treatment	0.0075 [0.0066]	0.0097 [0.0073]	0.0088 [0.0085]	0.0086 [0.0089]	0.0071 [0.0075]	0.0082 [0.0079]	0.0084 [0.0081]	0.0068 [0.0083]	0.0083 [0.0074]	0.0064 [0.0093]	
Sample size		178,219	148,737	116,009	84,032	51,534	178,219	148,737	116,009	84,032	51,534
R-squared		0.006	0.006	0.006	0.006	0.006	0.001	0.001	0.001	0.001	

Notes: Heteroskedasticity-robust standard errors clustered by day of birth are in parentheses. \*\*\*, \*\*, and \* indicate statistical significance at the 1, 5, and 10 percent level respectively. After is an indicator for individuals born after January 1. Treatment is an indicator that equals 1 for cohorts who experienced an education expansion.



**FIGURE 1: Graduates from Gymnasium Schools by Year of Graduation**

Notes: Figure 1 plots the number of students graduating from gymnasium between 1951 and 1971.

Source: Romanian Statistical Yearbook

Figure 2: Educational achievements in Romania by year of birth

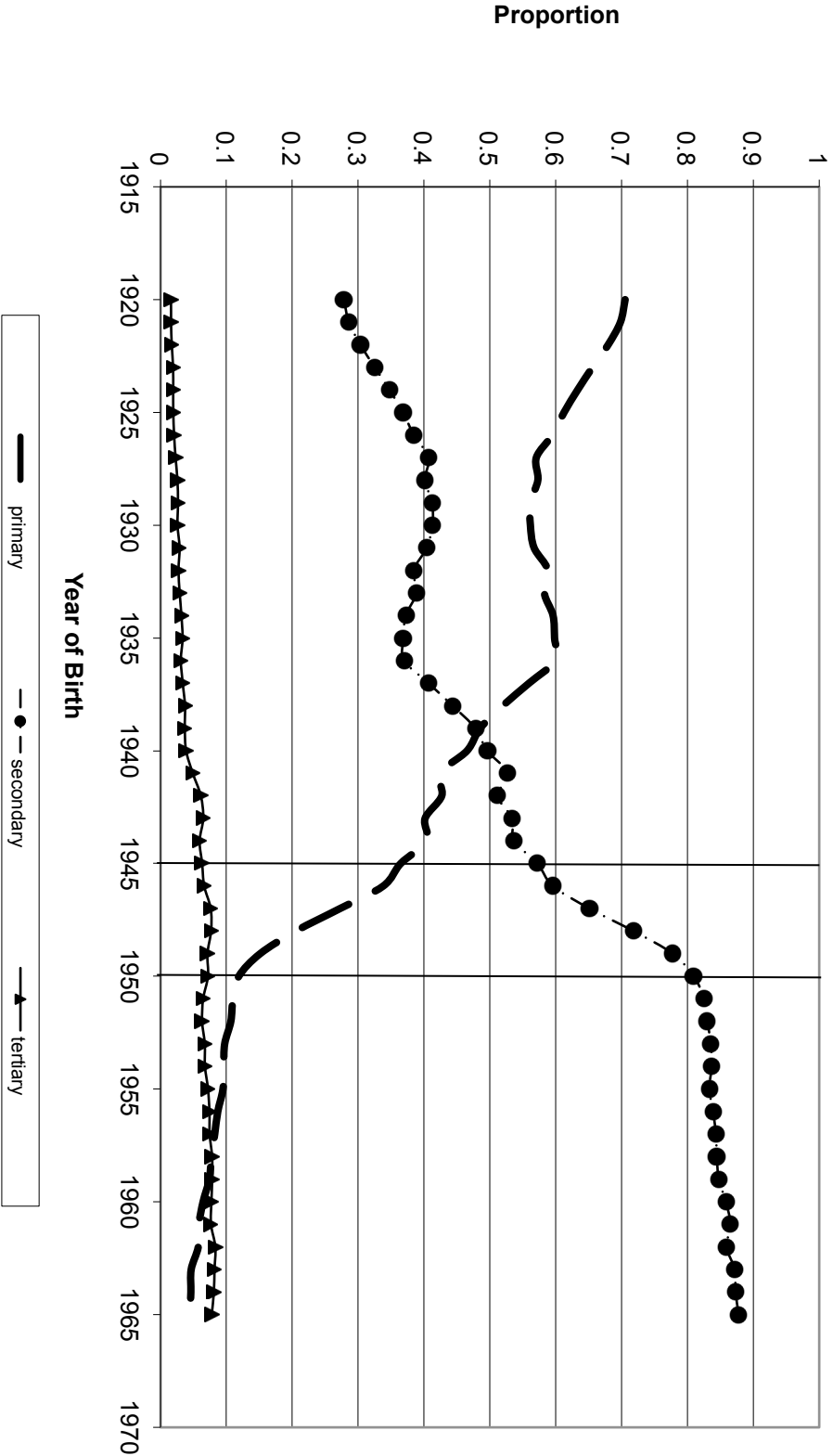
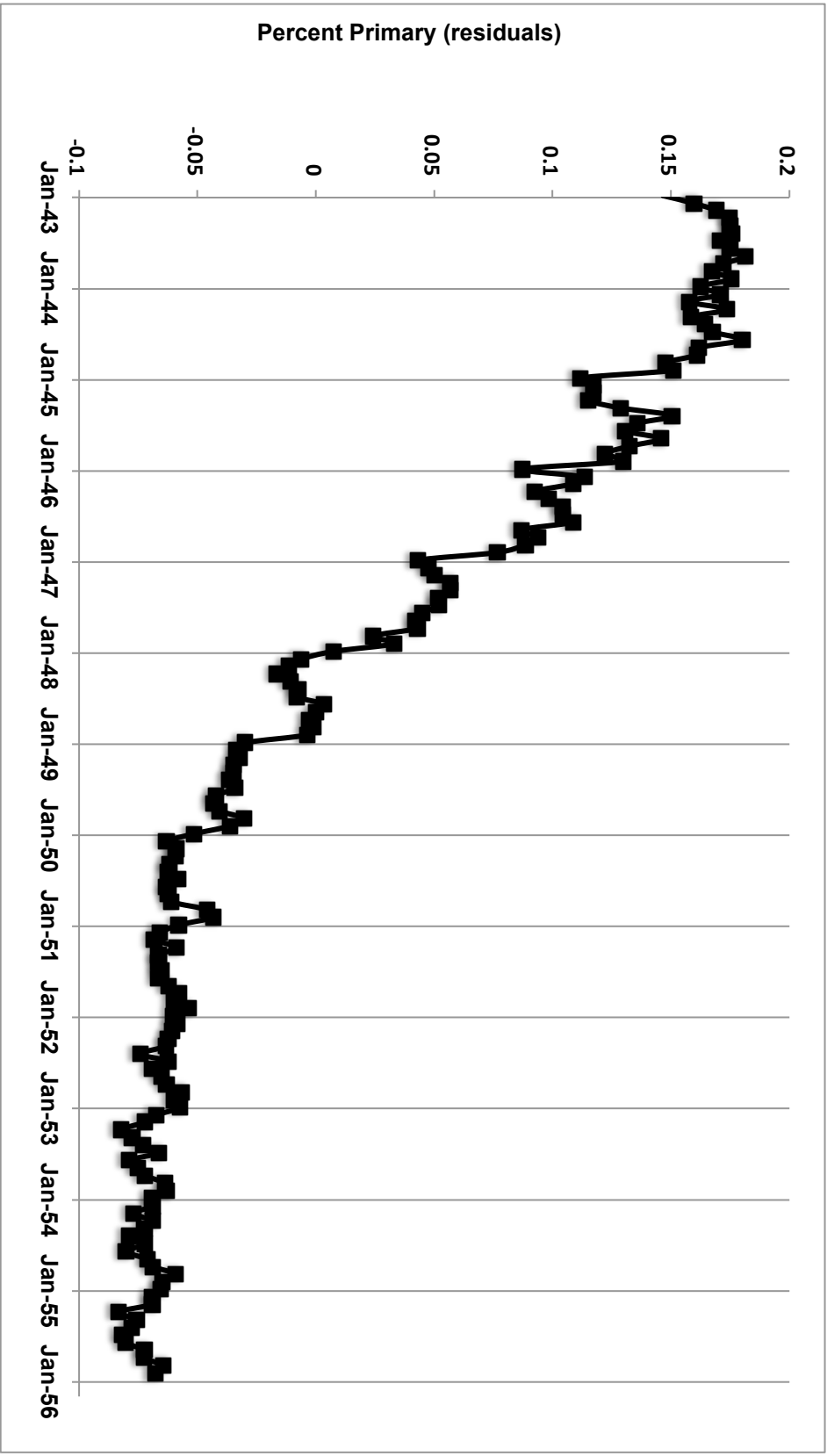
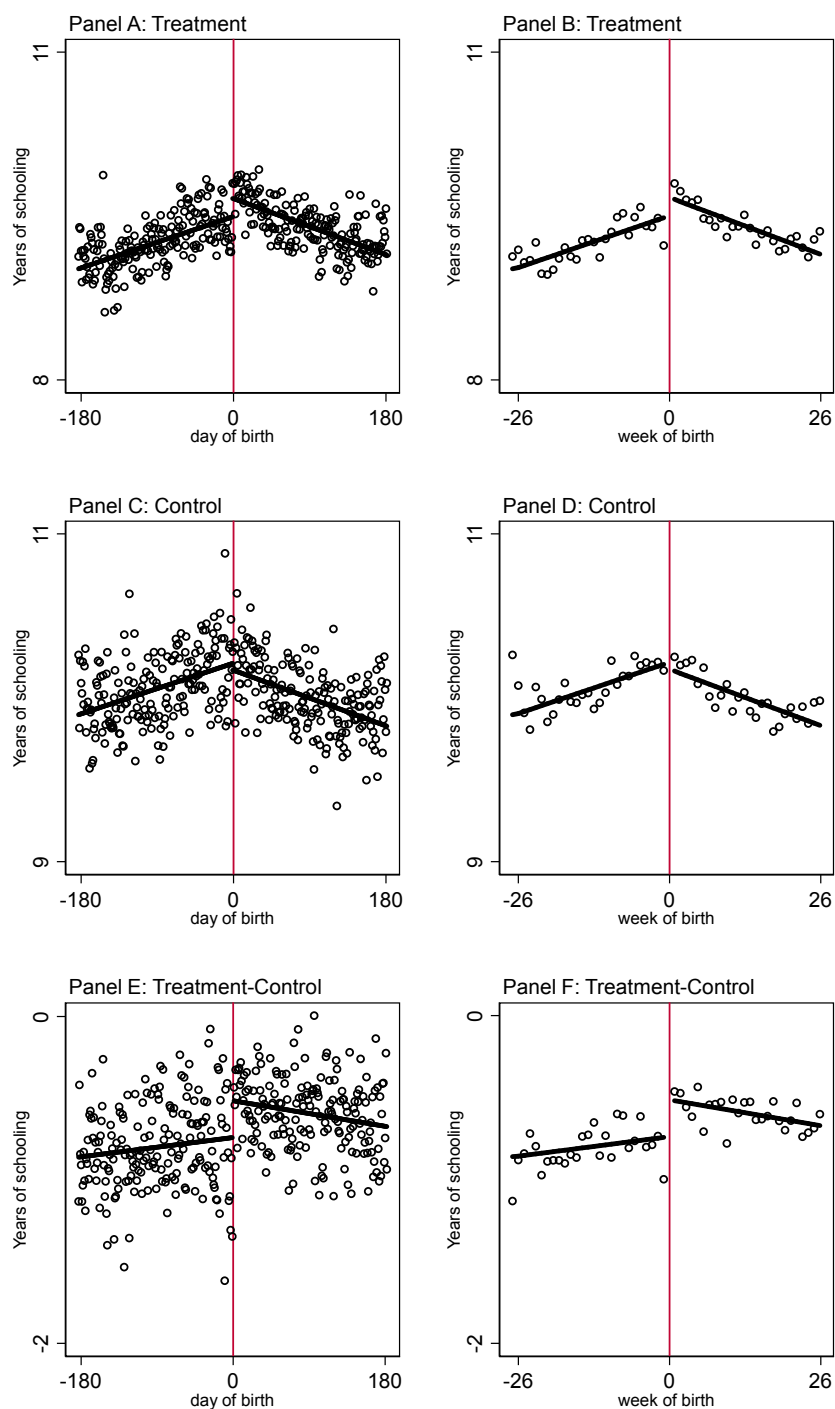


FIGURE 2: Educational achievements in Romania by year of birth  
Notes: Figure 2 plots the the highest educational attainment by year of birth for cohorts of individuals.  
Source: 1992 Romanian Census (PAU sample)



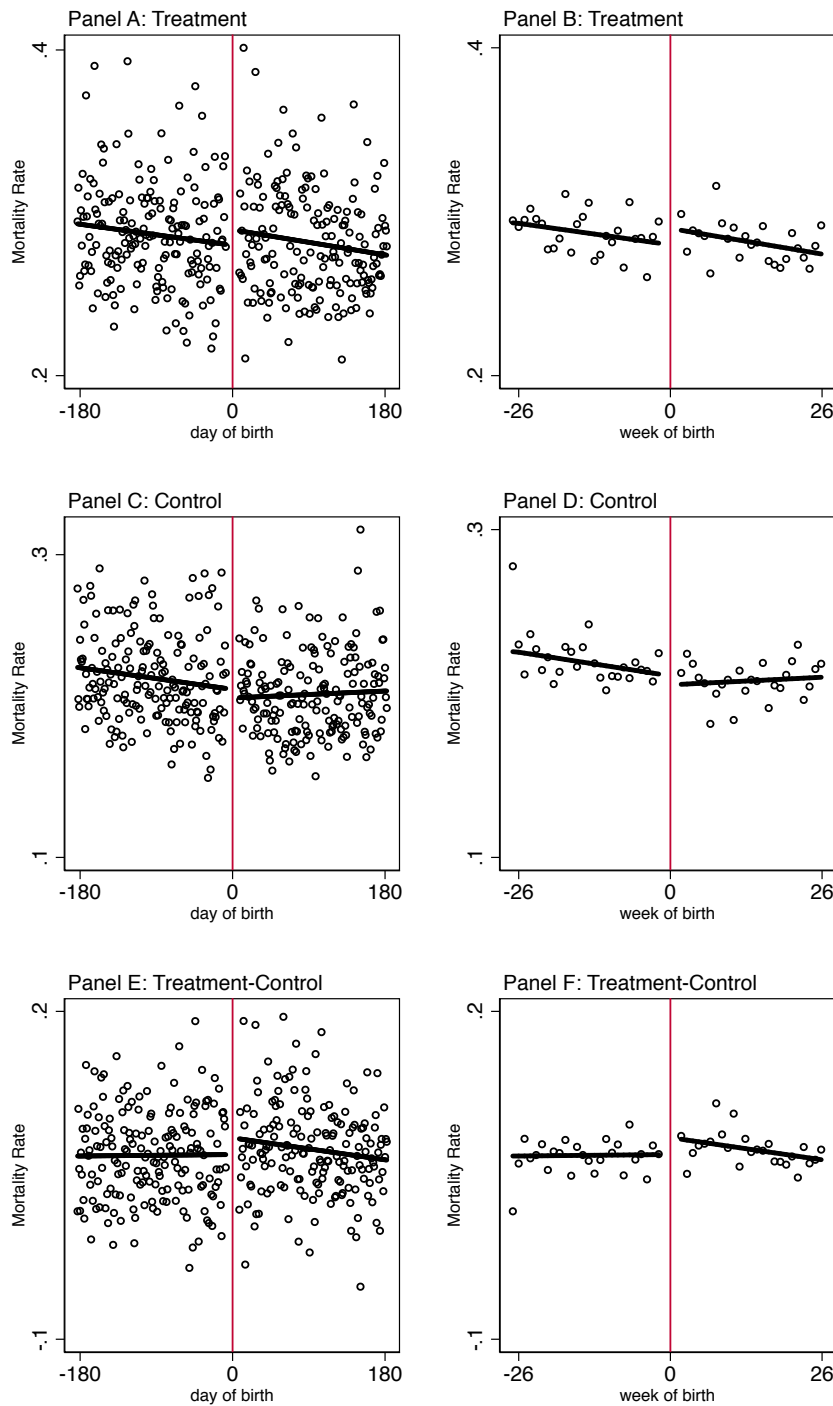
**FIGURE 3: Effect of educational expansion for cohorts born 1943-1955 by month of birth**

Notes: This figure plot the percent of individuals born between 1943 and 1955 who completed primary education by their month of birth, which are based on residuals. Source: 1992 Romanian Census (PAU sample)



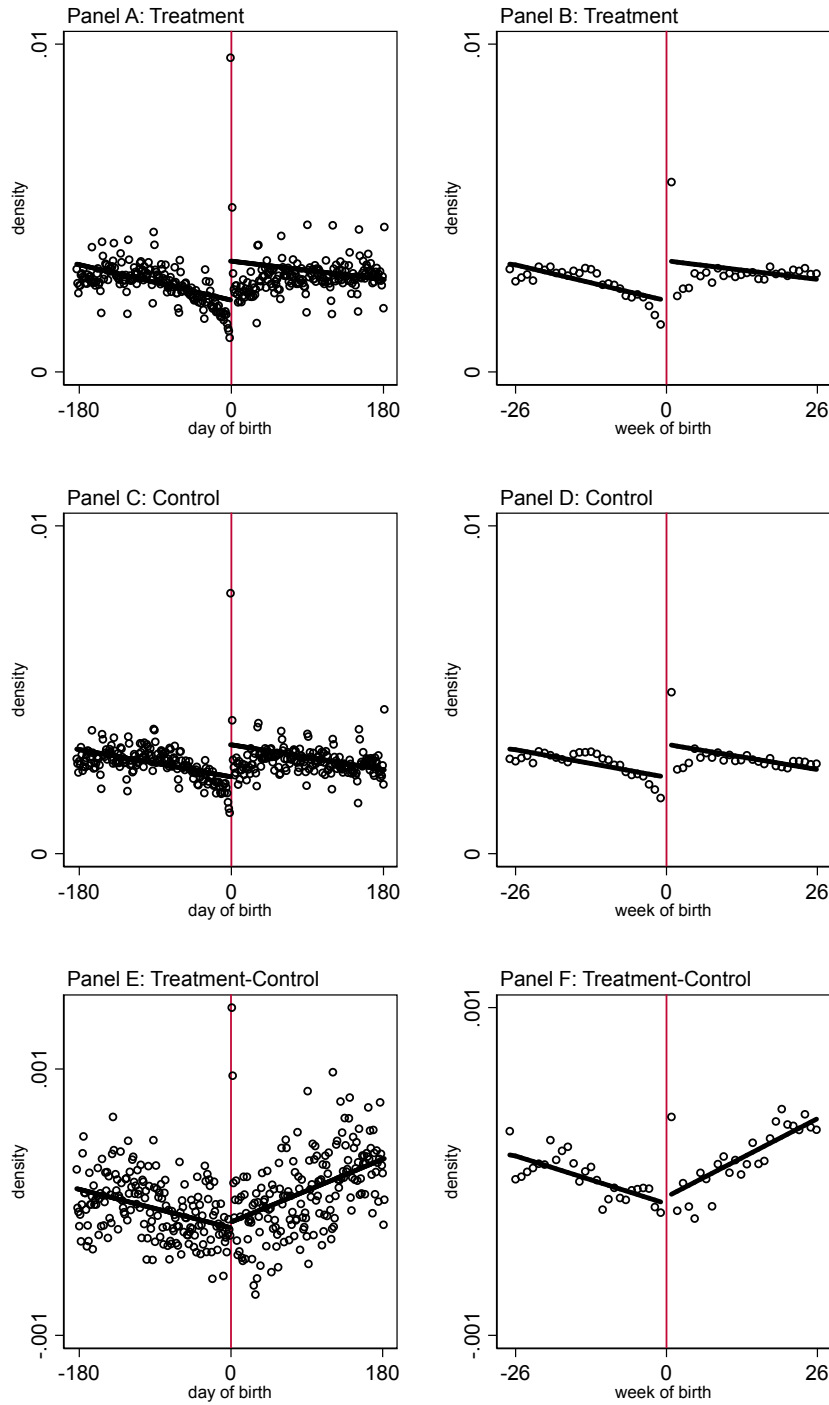
**FIGURE 4: Years of Schooling**

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1992 Romanian Census (PAU sample).



**FIGURE 5: Mortality Rate**

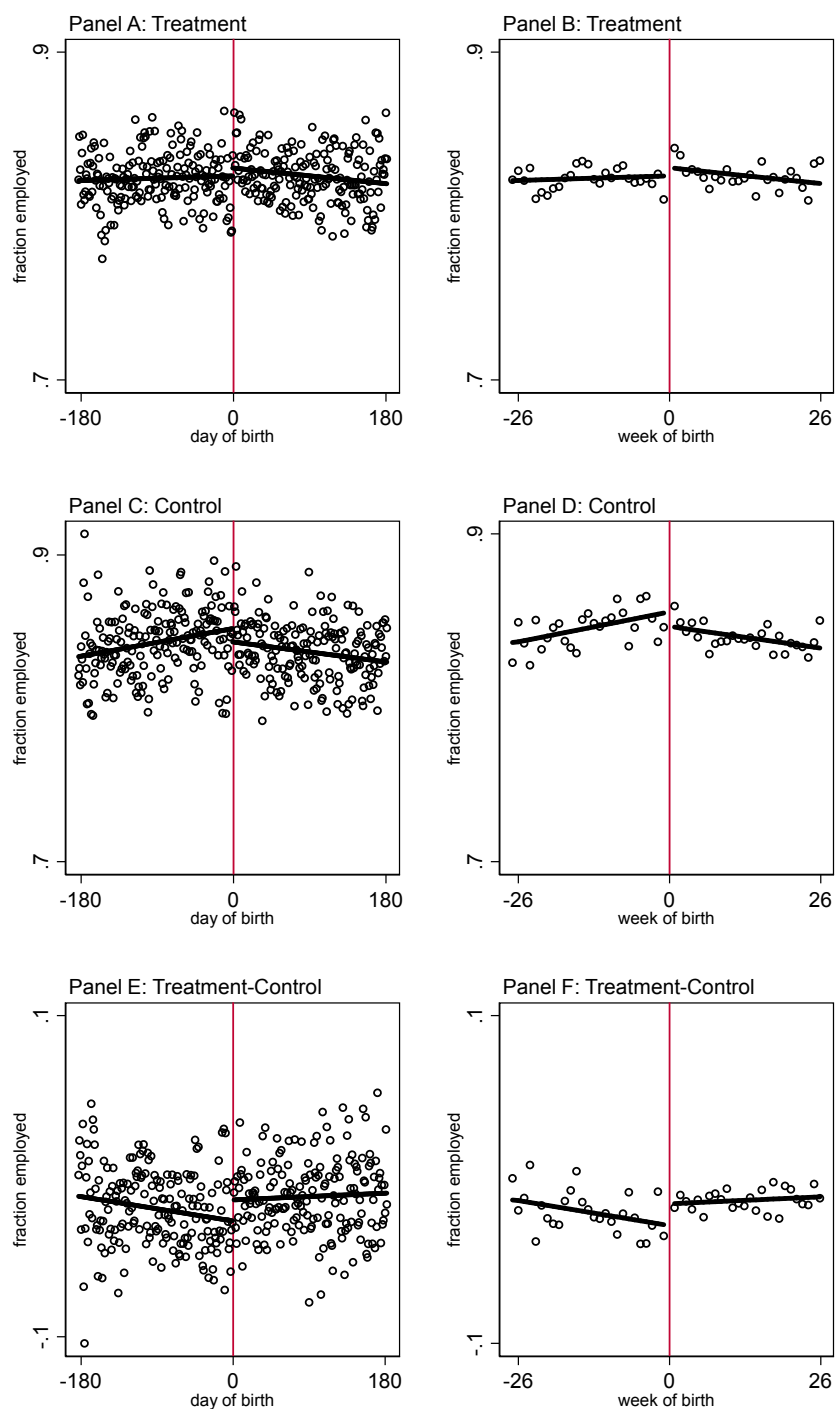
Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions.. Source: 1994-2011 Vital Statistics Mortality files



**APPENDIX FIGURE 1: Density check**

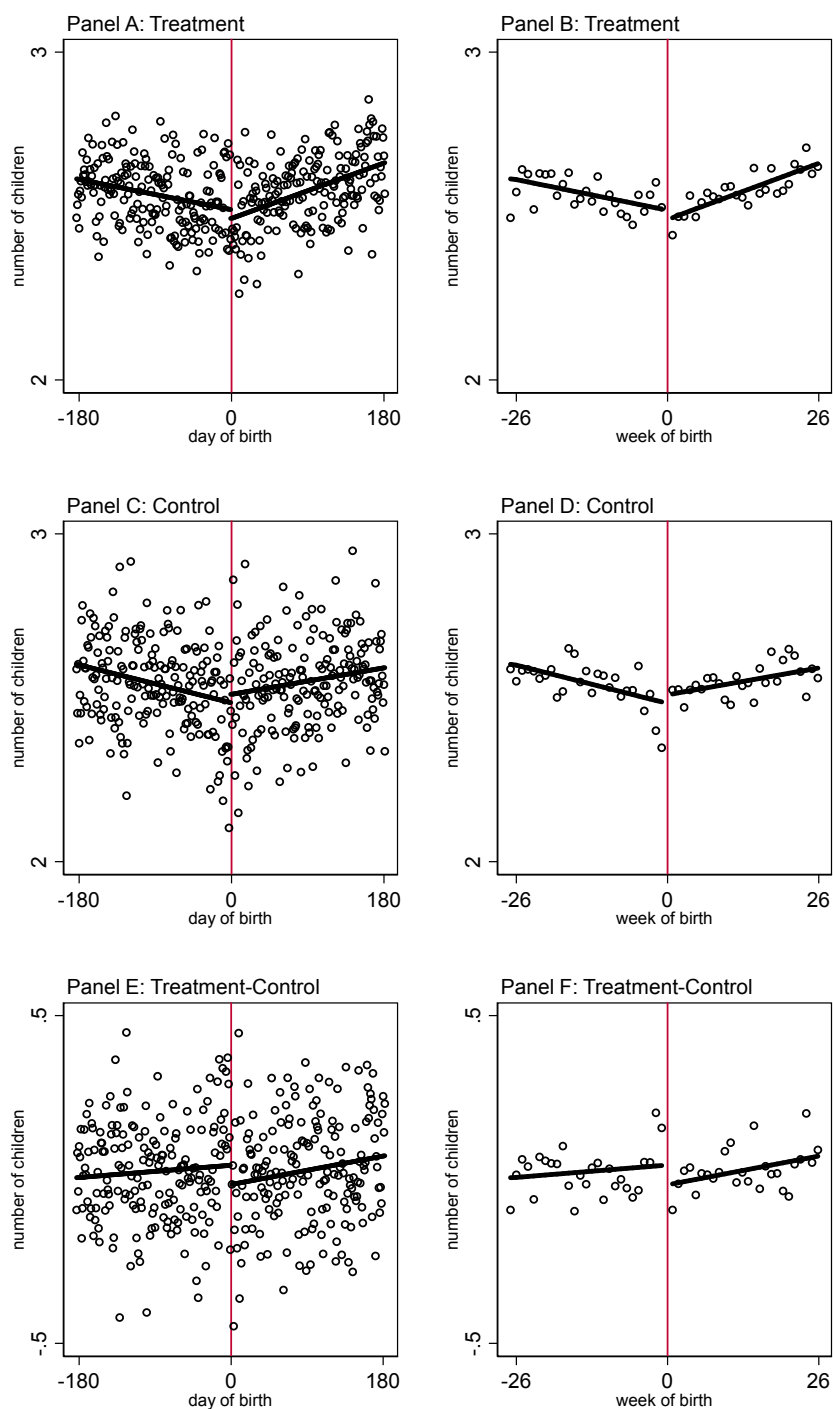
Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1992 Romanian Census (PAU sample).





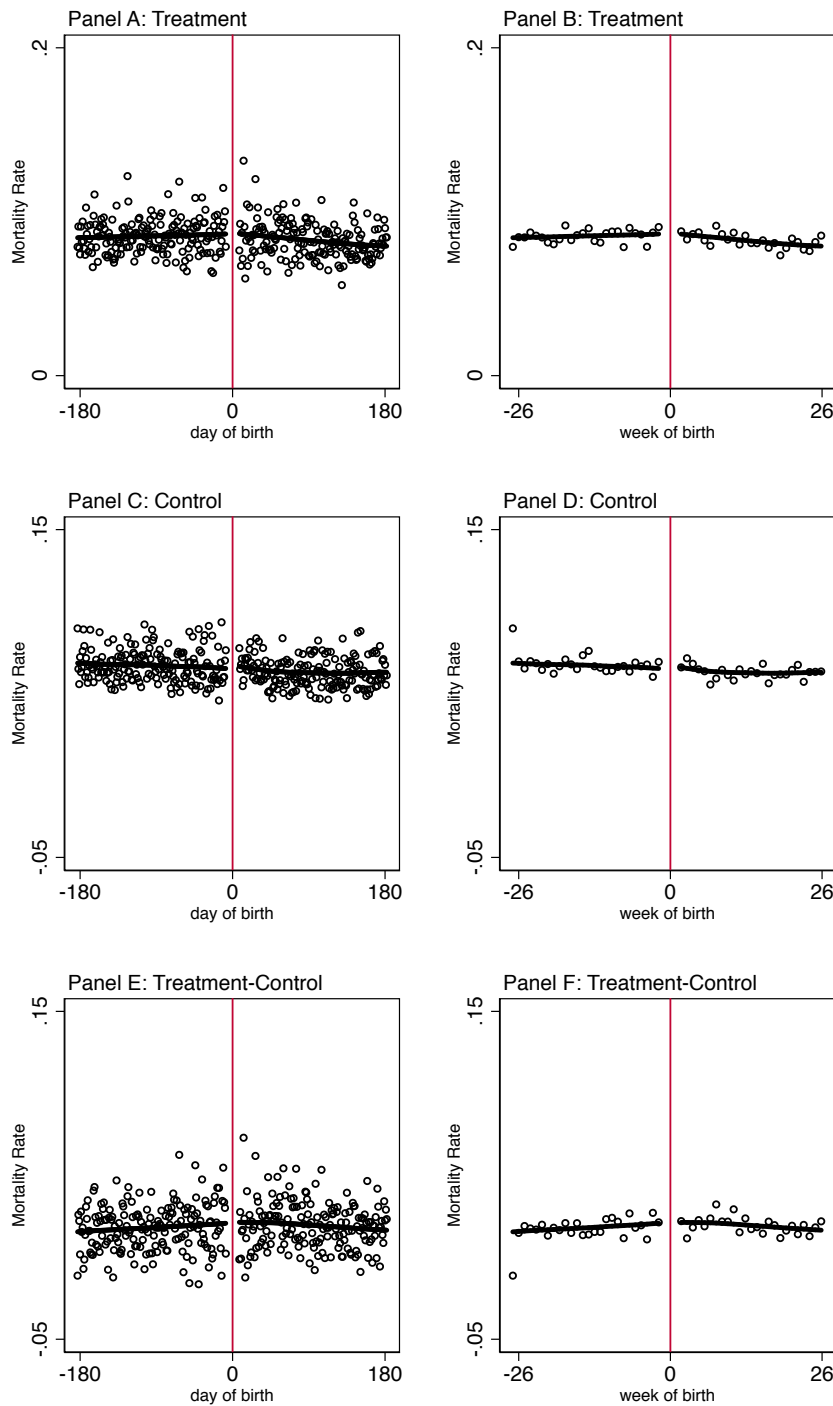
**APPENDIX FIGURE 2: Rate of Employment**

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1992 Romanian Census (PAU sample).



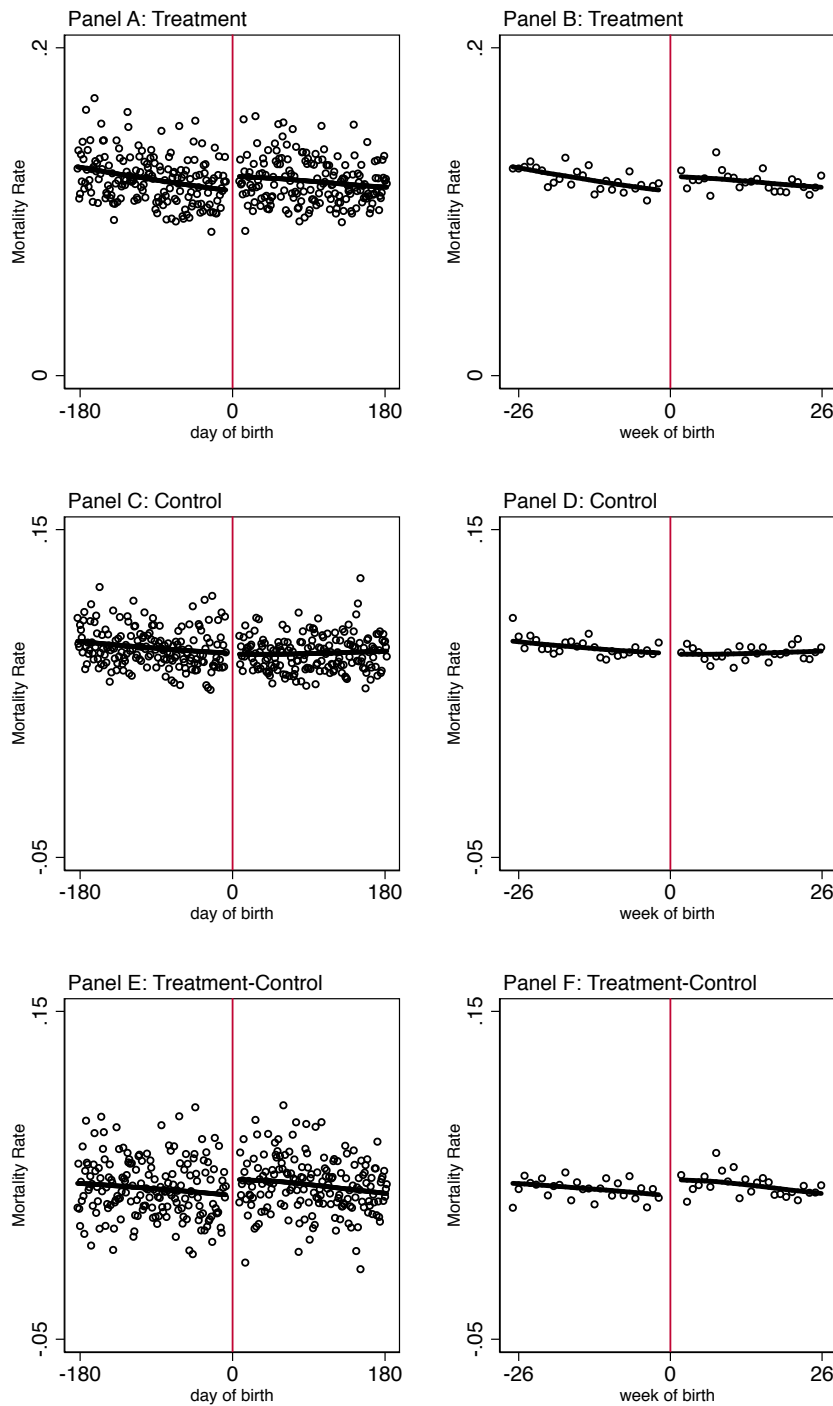
**APPENDIX FIGURE 3: Fertility**

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1992 Romanian Census (PAU sample).



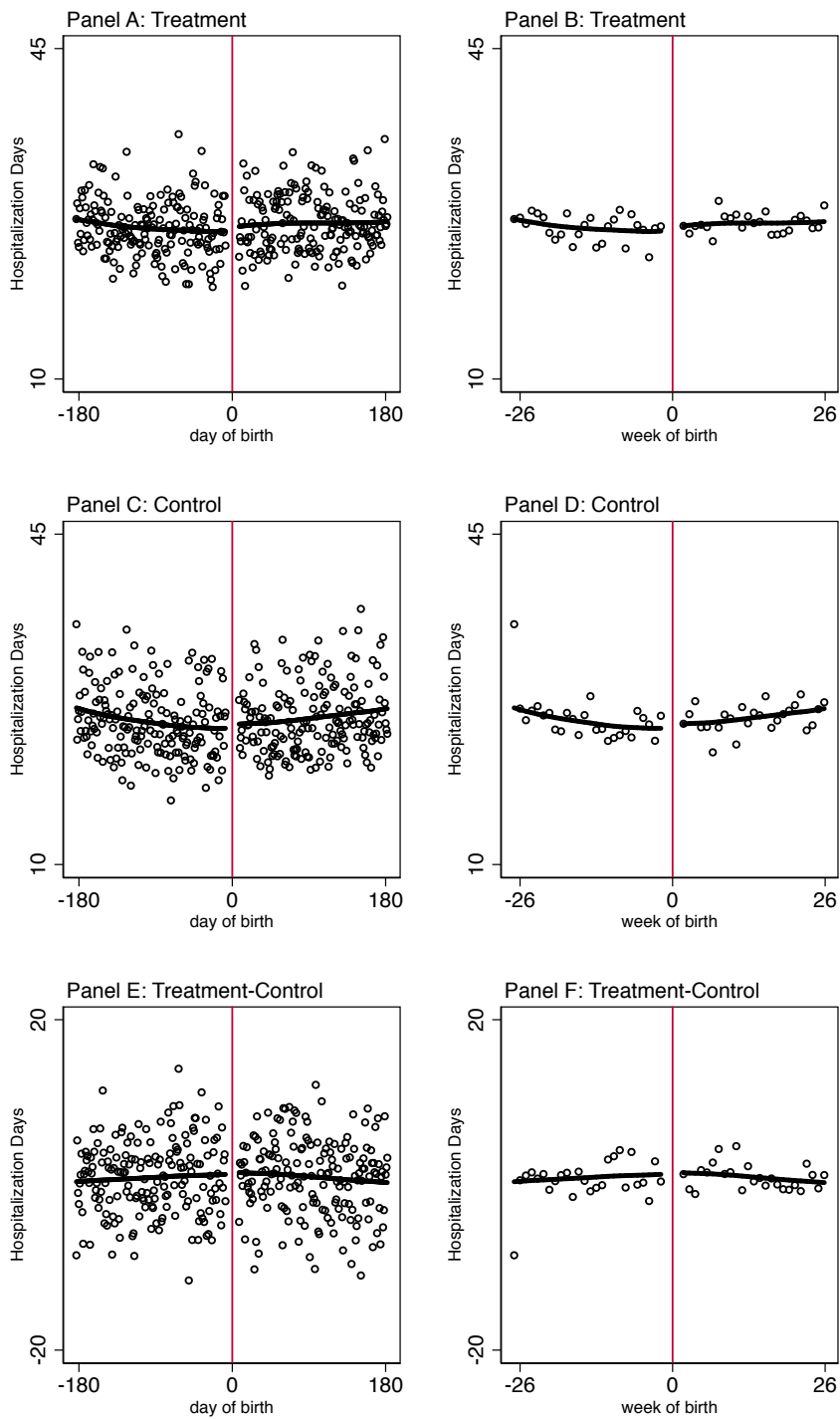
#### APPENDIX FIGURE 4: Mortality Rate due to Cancer

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1994-2011 Vital Statistics Mortality files



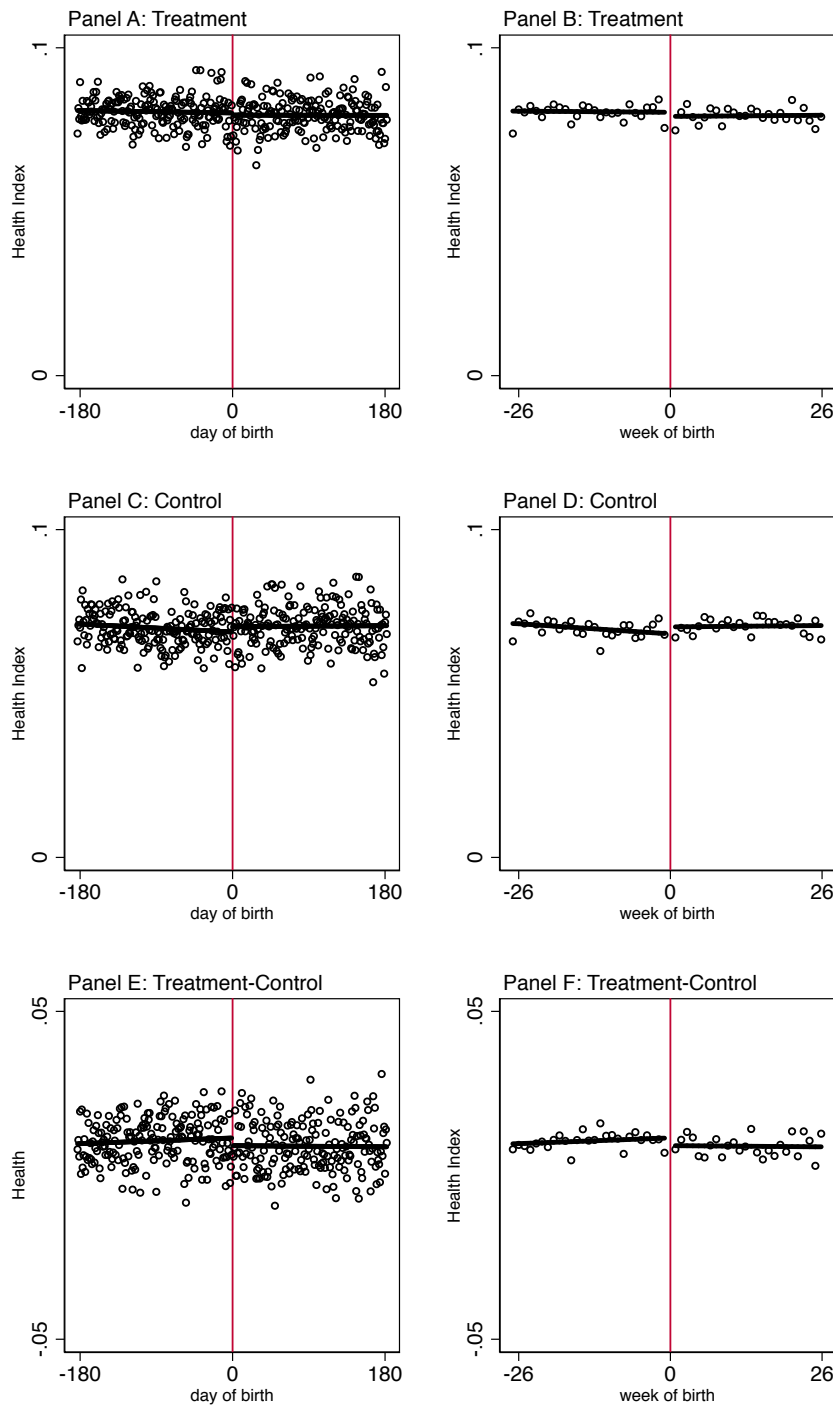
**APPENDIX FIGURE 5: Mortality Rate due to Circulatory Diseases**

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 1994-2011 Vital Statistics Mortality files



### APPENDIX FIGURE 6: Hospitalization Days

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions.



**APPENDIX FIGURE 7: Self-Reported Health Problems**

Notes: Panels A and B are restricted to individuals born in the treatment years (1944-1950). Panels C and D are restricted to individuals born in the control years (1950-1953). Panels E and F are restricted to individuals born in both treatment and control years (1944-1953). The open circles indicate the mean of the outcome by day of birth (panels A, C and E) or week of birth (panels B, D and F). The solid lines are based on linear spline regressions. Source: 2011 Romanian Census