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#### Abstract

Using representative linked employer-employee data for Germany, this paper analyzes short- and long-run differences in labor market performance of workers joining startups instead of incumbent firms. Applying entropy balancing and following individuals over ten years, we find huge and long-lasting drawbacks from entering a start-up in terms of wages, yearly income, and (un)employment. These disadvantages hold for all groups of workers and types of start-ups analyzed. Although our analysis of different subsequent career paths highlights important heterogeneities, it does not reveal any strategy through which workers joining start-ups can catch up with the income of similar workers entering incumbent firms.

Keywords: startups, young firms, wages, linked employer-employee data

JEL Classification: J31, J63, L26, M51

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#### 1. Introduction

The role of newly founded firms, so-called start-ups, in structural change and job creation is a highly disputed topic both in scientific and political debates (Shane 2009). There exists a broad empirical literature focusing on the quantity of jobs created (and destroyed) in new firms, across regions and at the aggregate level, mostly finding positive net effects of start-ups (see, e.g., Haltiwanger et al. 2013 for the U.S., Fritsch and Weyh 2006 for Germany, Criscuolo et al. 2014 for 18 countries, and the review by Block et al. 2018). In contrast, relatively few studies have analyzed the quality of these jobs from the viewpoint of the individual worker. Some of these studies suggest that job quality in start-ups may be questionable, but evidence so far is too scarce to make any definite statements (see the reviews by Block et al. 2018 and Nyström 2019). For workers – be they employed or unemployed – it is largely an open question whether joining a newly founded rather than an incumbent firm is advisable or not to achieve a sustainable (re-)integration into the labor market. Hence, the primary objective of this paper is to analyze empirically whether working at a start-up does pay off in the short and long run for individual workers. Are there temporary or persistent advantages and disadvantages in terms of remuneration and (un)employment prospects from entering a start-up rather than an incumbent firm?

Although the quality of jobs is a multi-dimensional concept that also includes work content and non-monetary benefits (Block et al. 2018), the employment and earnings prospects individuals face in start-ups surely play a major role. Workers entering a start-up rather than an incumbent firm may receive higher wages as a compensation for the higher failure risk of start-ups, but they also could initially face lower wages due to the financial constraints of their young employer operating at an inefficient scale (Brixy et al. 2007). In the latter case, working at a start-up could pay off in the long run if new firms survive and become more profitable (Nyström 2019). Wages in start-ups might even rise more steeply than in incumbent firms if flat hierarchies in expanding young firms mean that the initial workers are first in line to reach better-paid positions quickly (Fackler et al. 2019). Related, the greater variation in performed tasks and the expanded responsibility individuals typically experience in (small) startups may accelerate their career progression and earnings growth when moving to other, more mature firms (Sorenson et al. 2018). On the other hand, wage profiles could be steeper in incumbent firms if these are more likely to offer backloaded compensation schemes to their employees - a strategy that will be less credible for risky new firms (Schmieder 2013). Regarding un(employment) prospects, workers in newly founded firms face a

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high risk of involuntary job loss due to their employer's closure (Haltiwanger et al. 2013, Fackler et al. 2013). Hence, entering a start-up might be associated with a higher risk of unemployment and worse future labor market opportunities due to displacement and stigma effects (Sorenson et al. 2018). These brief considerations suggest that it is initially not clear whether entering a start-up as opposed to an incumbent firm will pay off for workers in the long run.

Previous research has primarily compared average wages in start-ups and incumbents or focused on differences in workers' entry wages at the point of being hired. The empirical evidence so far is ambiguous (see the reviews by Block et al. 2018 and Nyström 2019). While some papers show that wages are significantly lower in start-ups than in incumbent firms, *ceteris paribus* (e.g. Nyström and Elvung 2014, Fackler et al. 2019), others find a positive wage differential, in particular for very successful start-ups (e.g., Schmieder 2013, Ouimet and Zarutskie 2014, Burton et al. 2018). Brixy et al. (2007) identify a negative wage differential that becomes smaller over time, but they only have data at the level of establishments and not of workers. Babina et al. (2019) report a pay penalty at young firms that turns into a small pay premium after controlling for various dimensions of worker and firm heterogeneity.

Very few papers have been able to follow workers and their wages over time.<sup>1</sup> Analyzing linked employer-employee data from Britain, Adrjan (2018) finds that young firms pay slightly higher wages to new hires, but subsequent wage growth is steeper at mature firms. He demonstrates that this finding holds both within continuing employment relationships and for individuals who change jobs, but he is not able to further analyze workers' (un)employment trajectories. Using Danish registry data, Sorenson et al. (2018) show that individuals who join young firms (i.e., less than four years old) earn substantially less than the employees of large, mature firms over the subsequent ten years, and these earnings disparities are not found to diminish over time. Both above mentioned studies, however, focus only on remuneration as the sole indicator for labor market success. In her recent review article, Nyström (2019, p. 10) concludes that "there is a clear scarcity of research regarding the long-term wage

<sup>&</sup>lt;sup>1</sup> In their seminal work, where they find no robust and statistically significant wage differentials between workers in young and old firms, Brown and Medoff (2003) try to approach the long-term effects by controlling for tenure in their cross-sectional analysis. Babina et al. (2019) estimate earnings-tenure profiles for workers in young firms over the first three years and find evidence for flatter profiles in young firms when controlling for various dimensions of worker and firm heterogeneity.

trajectories of employees in entrepreneurial firms." In addition, there is a lack of studies that look at the long-term (un)employment trajectories of individuals.<sup>2</sup>

Our paper contributes to this small literature and goes beyond previous studies in various ways. First, when asking whether it pays off to enter a start-up rather than an incumbent establishment, we not solely focus on wages but also consider other indicators of labor market success such as days in employment and unemployment benefit receipt. Second, using a large, representative linked employer-employee data set for Germany, we follow individuals joining a start-up over ten years and analyze whether there are differences in wages and (un)employment compared to similar individuals who have entered incumbent firms. To ensure comparability of the two groups of workers, we apply entropy balancing (Hainmueller 2012). We then examine whether the remaining differences are only temporary or long-lasting and whether they vary for different groups of workers. Third, we further add to the literature by investigating various potential explanations for the observed short- and long-term differences, such as joining successful vs. failing start-ups or pursuing different subsequent employment paths (like staying or leaving the establishment). The upshot of our empirical analysis is that there are huge and long-lasting drawbacks from entering a start-up rather than an incumbent establishment, both in terms of income and employment and for all groups of workers and types of start-ups analyzed.

The remainder of the paper is organized as follows: Section 2 explains our data and provides descriptive evidence on the composition of workers entering either a new or an incumbent firm. The methods and results of our econometric analyses are presented and discussed in Section 3. Section 4 concludes.

# 2. Data and descriptive evidence

To analyze the different labor market prospects of workers entering either a start-up or an incumbent firm, we use an extensive linked employer-employee data set for Germany based on social security notifications, which is provided by the Institute for Employment Research (IAB). Our data set combines worker level information from the Integrated Employment Biographies (IEB) and establishment level information from the Establishment History Panel (BHP).

<sup>&</sup>lt;sup>2</sup> A partial exception is the study by Schnabel et al. (2011) for Germany that compares the employment stability of one cohort of workers in newly founded and incumbent firms over a period of six years.

Detailed data on labor market participants is collected in the IEB, which provides daily information on employment relationships for all workers subject to social security notifications, as well as periods of benefit receipt, registered job search, and participation in active labor market programs from 1975 to 2014 for Western Germany.<sup>3</sup> Since 1992, Eastern Germany is included in the data as well, and from 1999 onwards, information on marginally employed workers is collected, too. Additionally, the IEB contains individual characteristics such as age, gender, education, and nationality.<sup>4</sup>

Yearly information on all German establishments with at least one worker subject to social security contributions is contained in the BHP, including size, sector, location, and workforce composition as of June 30 of a given year.<sup>5</sup> Crucial for our analysis of newly founded establishments, the BHP also contains information on worker flows (Hethey-Maier and Schmieder 2013). In order to distinguish whether a new establishment identifier in the data refers to a truly new entry or is caused by mergers, acquisitions, or other changes of the identification number, worker flows are used to identify which fraction of a new establishment. We restrict our analysis to newly founded establishments defined by Hethey-Maier and Schmieder (2013) as "new (small)" or "new (med & big)", implying that the establishment either employs not more than three workers in its first year of business, or, if larger, less than 30% of its initial workforce have worked together under a common establishment identifier in the previous year.

Moreover, it must be noted that establishments in the BHP are defined as local production units, which do not necessarily correspond to firms as legal entities. Since we intend to focus our analysis on the foundation of new, independent firms instead of branch openings of multi-plant firms, we exclude establishments with more than 20 employees in their first year of business. We evaluate the success of this procedure in reducing the number of branch openings by using information from the IAB Establishment Panel, a yearly survey of approximately 16,000 German establishments.<sup>6</sup> Since the Establishment Panel includes information on single- and multi-plant firms, we can link this information with those establishments from the BHP

<sup>&</sup>lt;sup>3</sup> This implies that the IEB only includes information on hired employees. The founders of the firms are not listed in the data, since they are not subject to social security contributions.

<sup>&</sup>lt;sup>4</sup> For more information on the IEB, see Antoni et al. (2016) who provide a description of the Sample of the Integrated Labour Market Biographies (SIAB), a 2% random sample from the IEB.

<sup>&</sup>lt;sup>5</sup> For detailed information on the BHP, see Schmucker et al. (2016).

<sup>&</sup>lt;sup>6</sup> For further information on the IAB Establishment Panel see Ellguth et al. (2014). We do not use the IAB Establishment Panel in our main analysis, even though it includes some additional information at the firm level, because the number of young establishments in the data set is rather small and typically establishments in their very first year of existence are not included in the survey at all.

that we classify as start-ups as described above and that meet further sample restrictions described below. It can be shown that circa 94% of the establishments we define as start-ups are independent new firms, while only 6% are branch openings of existing entities.

The sample of start-ups that is used for our analyses consists of a 10% random draw of all establishments newly founded in the years 2000 to 2004, only focusing on establishments in their very first year of business. We then link information from the IEB on all newly hired workers in the respective year, i.e., workers that have not been working with the same employer in the previous year. Since workers' employment biographies are available until 2014, this allows us to follow each cohort of workers (and firms) over ten subsequent years. The control group of incumbent establishments is constructed by drawing a 5% sample of all establishments existing during that period. For each cohort of workers, we keep only those entering establishments which are five years or older.<sup>7</sup> In both groups, we exclude establishments in agriculture, energy and mining, and in the public sector. We further exclude workers younger than 18 and older than 50 at the time of being hired, as well as apprentices.

Table 1 gives an overview over the establishments in our final sample. New establishments in their first year of business are, on average, much smaller than incumbents, and they are significantly more often found in the tertiary sector. The two groups of establishments also differ in terms of their workforce composition, concerning the share of women, full-time workers, and marginally employed workers among their employees.

To provide an overview of the composition of workers entering new and incumbent establishments, respectively, we present selected individual characteristics at the point of entry in Table 2. We see that the two groups differ significantly in almost all variables presented. Workers entering new establishments are more often women and they are on average older than the control group. They are more often medium-qualified, while a higher share of workers entering incumbent establishments is either low-qualified, i.e., having no degree at all, or high-qualified, i.e., graduated from university. Moreover, workers taking up a job in a start-up are less often of German nationality, have less frequently performed a job-to-job transition<sup>8</sup> and are less often hired in a part-time job.

<sup>&</sup>lt;sup>7</sup> The threshold of five years might appear arbitrary, but Brixy et al. (2006) show that after the first five years of business, differences in wage levels and working conditions between new and incumbent firms become insignificant.

<sup>&</sup>lt;sup>8</sup> Following Fackler et al. (2019), we define job-to-job transitions as recruitments where individuals left their previous job not more than 90 days before joining the respective establishment, hence allowing for

In terms of years of working experience, we find no significant differences, while workers entering new establishments have previously spent more time in benefit receipt.<sup>9</sup> Moreover, individuals entering start-ups have had more previous employers, which points towards more stable employment biographies in the control group. All these differences in the sample composition might affect the labor market success of the two groups of workers. Our goal in the following empirical analysis is to study workers' employment trajectories in the long run and to investigate whether various indicators of labor market success differ between workers entering either a start-up or an incumbent, thereby conditioning on a broad range of individual and firm characteristics.

#### 3. Empirical analysis

## 3.1 Econometric approach

To account for differences in the composition of the groups of workers entering startups vis-à-vis incumbents, we apply entropy balancing (Hainmueller 2012, see Hainmueller and Xu 2013 for a description of the respective Stata ado-file *ebalance*). This method allows us to directly impose the first and second moments, i.e., means and variances, of a large set of covariates to be perfectly balanced among both groups. Without having to postulate any further assumptions, entropy balancing reweights observations to match the respective balance constraints by deviating as little as possible from the initial weights. By directly focusing on covariate balance, entropy balancing improves on related methods such as propensity score matching, which often depend on manual adjustment of the weighting scheme and repetitive balance checking and therefore frequently fail to balance all covariates perfectly. Moreover, while matching approaches often discard less comparable individuals in the control group, entropy balancing retains all relevant information by assigning weights smoothly to all observations in the data (Hainmueller and Xu 2013).<sup>10</sup> In our case, we aim to compare two groups of individuals with the same preconditions when joining an

a short period of frictional unemployment. If workers left their previous job more than 90 days ago and in the meantime were registered as a job seeker, received benefit payments, participated in labor market programs or were not observed in the data, they are not defined as transitioning from employment.

<sup>&</sup>lt;sup>9</sup> Note that working experience and benefit receipt are truncated at eight years because eastern German workers only joined the data in 1992. Hence, for the cohort of workers hired in 2000 we can only observe eight years of their employment histories.

<sup>&</sup>lt;sup>10</sup> To check whether our results depend on the empirical method chosen, we additionally run a robustness test where we substitute entropy balancing with propensity score matching. Results are almost identical to the main outcomes discussed below and are available upon request.

establishment, so that diverging trajectories in labor market performance in the subsequent years can be more credibly ascribed to entering either a new or an incumbent establishment.

More specifically, we require observations in the control group to be reweighted so that means and variances of the workers' year of entry, sex, age, qualification and German nationality equal those of the group of workers entering a start-up, since all these characteristics typically account for differences in individual career paths and wages. We also balance the two groups in terms of preceding employment status, indicating whether an individual either has performed a job-to-job transition or has come from unemployment or outside the labor market, and in terms of total previous years of experience and years of benefit receipt. In addition to these variables that might affect workers' labor market opportunities, we include the number of former employers in the balancing procedure to capture previous employment stability. Moreover, we also impose the two groups to be balanced concerning the new job's part-time status and occupation, as well as the sector and labor market region of the establishment.<sup>11</sup> We then investigate individuals' labor market performance over time in terms of yearly income, average daily full-time earnings<sup>12</sup>, days in employment, and days of benefit receipt<sup>13</sup> in the reweighted sample for the next ten years following workers' entry in the respective establishment. To compare these indicators between the two groups of workers, we run an OLS regression in the balanced sample,

$$Y_{it} = \sum_{t} \beta_t T_t + \sum_{t} \gamma_t S U_i T_t + v_{it}$$

where  $Y_{it}$  determines the labor market outcome of interest for individual *i* in year *t* and *T* indicates a set of relative time dummies, ranging from zero in the year in which the individual newly enters the establishment up to year 10. Additionally, these time dummies are interacted with a start-up indicator  $SU_i$  that is equal to one if the worker

<sup>&</sup>lt;sup>11</sup> We categorize occupations according to Blossfeld (1987). Sectors are defined at the two-digit level and labor market regions are classified on the basis of workers' commuting patterns according to Kropp and Schwengler (2011). We do not include establishment size in our balancing procedure because comparing small start-ups with similarly small incumbents may be misleading. A small start-up might grow quickly in its first years, whereas as small established firm potentially signals that it has not been sufficiently successful to grow. Nevertheless, we also include a robustness check making start-ups and incumbents more comparable in size, which is discussed in Section 3.3.

<sup>&</sup>lt;sup>12</sup> Note that our indicator for wages, average daily earnings, is defined conditional on full-time employment. Since our data does not contain information on working hours, we are not able to calculate hourly wages. Hence, part-time workers are excluded from the analysis of wages to reduce heterogeneities in working hours.

<sup>&</sup>lt;sup>13</sup> Benefit receipt refers to unemployment insurance benefits (*Arbeitslosengeld I*) and means-tested social assistance (*Arbeitslosengeld II*).

had entered a new establishment and zero for workers who had joined an incumbent firm at the beginning of the observation period. The coefficient  $\gamma_t$  therefore shows the difference in the performance of the two balanced groups of workers for each year.

Our empirical approach allows us to render the two groups of workers perfectly comparable among a broad set of observable characteristics. However, it must be acknowledged that there may be further dissimilarities between the individuals which we are not able to capture with our identification strategy, but which could affect their future career paths as well. For example, workers entering start-ups might be less risk averse than workers who choose to work for an incumbent. One might also imagine workers joining new firms to have a stronger preference for non-monetary aspects of a job, such as flat hierarchies, more autonomy or more diverse tasks which are often associated with working at a start-up (Sorenson et al. 2018). These characteristics may also play a role in workers' future career decisions and affect their success in terms of wages and employment prospects.

Since these (and other) unobservable differences could bias our estimate of differences in labor market performance, we additionally apply a robustness check in which we include workers' labor market outcomes of the three preceding years in our balancing procedure. By controlling for income, as well as days in employment, full-time employment, and benefit receipt, of the three years prior to entering the establishment, we abstract from any unobservable differences between the two groups of workers that had affected their labor market trajectories before our observation period.<sup>14</sup> Results of this robustness test will also be discussed in the following. However, we do not control for labor market outcomes in preceding years in our main analysis since introducing the additional variables into entropy balancing would force us to discard all labor market entrants from our sample.

#### 3.2 Results

The labor market trajectories of workers entering either a start-up or an incumbent establishment, both before and after entropy balancing, are presented in Figure 1. A first look already reveals that workers who joined a start-up in year zero perform worse in terms of all outcome variables over the whole observation period. Even though

<sup>&</sup>lt;sup>14</sup> We do not control for average daily full-time wages (conditional on full-time employment) because we would have to discard all individuals who spent one or more of the three preceding years in part-time employment or non-employment.

entropy balancing strongly reduces the gap between the two groups of workers, pointing towards negative selection into start-ups, the overall patterns remain stable.

Taking a closer look at each labor market outcome, one can see that workers entering a start-up already have lower yearly incomes in the year of entry, even after balancing.<sup>15</sup> This gap seems to widen slightly in the first years and then remains very persistent, without any indication that workers who initially entered a new establishment catch up to the control group. It should be noted that our indicator for yearly income captures two aspects, an employed worker's wage and (periods of) nonemployment in the respective year, the latter being assigned zero earnings. We therefore disentangle the two aspects by looking separately at wages (conditional on full-time employment) and days in employment. Focusing on average daily full-time earnings first, Figure 1b shows lower wages for workers in start-ups already in year zero, and the difference to the balanced group of workers entering incumbents hardly changes during the ten subsequent years. In terms of days in employment as well as days of benefit receipt, there is more variation over time. While differences in the year of entry are comparably small, the gap between the two groups widens considerably in the following two years, potentially picking up the effect of higher failure rates among start-ups. There seems to be some convergence in terms of days in employment, but workers who initially entered a start-up still perform worse than the control group even after ten years.<sup>16</sup>

In order to assess the differences in labor market performance and their statistical significance, Figure 2 shows the estimation results of the OLS regression described above. More specifically, the lines indicate the magnitude of the coefficients  $\gamma_t$  and the respective 95% confidence intervals for estimations in the unbalanced and the balanced sample. Our results confirm that workers entering a start-up perform significantly worse than the control group over the subsequent 10 years. Even after entropy balancing, they earn about  $\in$  4,000 (or approximately 20 percent) less yearly income from the second year onwards compared to workers who joined an incumbent firm, and this gap remains stable until the end of our observation period. Two factors contribute to this difference in yearly income: one is the persistently lower wages of approximately  $\in$  10 (roughly 15 percent) less per day and the other is the continuously

<sup>&</sup>lt;sup>15</sup> This finding is consistent with the study by Fackler et al. (2019), reporting that individuals receive lower entry wages when joining start-ups, *ceteris paribus*.

<sup>&</sup>lt;sup>16</sup> That Figure 1 shows less convergence in benefit receipt than in employment might reflect that the first measure also comprises workers who are back in employment, but whose wages are so low that they receive additional benefit payments.

lower probability of being employed. After two years, workers in a new establishment spend almost 20 days less in employment per year than their peers in incumbents, and while this gap is slightly reduced over the next years, differences remain highly significant throughout the observation period. The fact that there is also a strong increase in days of benefit receipt compared to the control group over the first two years after entry suggests that these workers usually do not have other income sources compensating employment losses.

As discussed above, one might be skeptical whether our empirical approach is successful in reducing all differences between the two groups of workers, since in entropy balancing we cannot control for unobservable characteristics such as ambition or risk aversion. We therefore present the results of a robustness check, in which we include indicators of labor market performance in the three preceding years in the balancing procedure, in Figure A1. It can be shown that although there are no remaining differences in terms of labor market success in the years -3 to -1 after reweighting the two groups, there are still substantial differences in labor market performance after entering the respective establishment, thus confirming the findings of our main specification.<sup>17</sup>

Moreover, we estimate an additional robustness check where we restrict our analysis to workers who enter an establishment with a maximum of 20 employees, to make both groups more comparable with respect to establishment size. In our preferred specification, we do not control for establishment size, since comparing small start-ups only to a group of similarly small incumbents may be misleading. While a small start-up might grow quickly in its first years of business, an established firm of comparable size potentially signals that it has been not so successful so far and therefore did not expand. Therefore, our main insights might also be driven by differences in establishment size coming along with establishment age. The results of this robustness check (Figure A2) show that even after making the two groups more comparable in terms of establishment size, we still find significant and persistent drawbacks from joining a start-up. While the differences in earnings shrink by more than half compared

<sup>&</sup>lt;sup>17</sup> Note that for this robustness test, we have to exclude all individuals with missing information on labor market performance for one or more of the three preceding years. To test whether this smaller sample differs strongly from our main sample in terms of subsequent labor market performance, we rerun the original balancing procedure (without controlling for previous labor market success) in this subsample and find that results are in line with our main results.

to the results of our preferred specification, differences in terms of employment prospects are still similar in size.

We further investigate whether our results also hold for the high-tech sector, which is often associated with especially successful and quickly growing young firms. Results for workers in manufacturing sectors with high research intensity, as defined for Germany by Gehrke et al. (2013), show again that workers joining start-ups experience worse subsequent labor market trajectories than workers entering incumbents (see Figure A3).

To sum up, our main results imply that workers entering a start-up suffer from severe and long-lasting drawbacks in terms of earnings and employment prospects, compared to workers joining an incumbent establishment instead. To analyze whether these insights hold for various subgroups of workers, we perform entropy balancing in subgroups defined by gender, age, qualification, and previous employment status. We further investigate whether our insights also apply to different percentiles of the income distribution. The respective regression results for yearly income as a summary measure for wages and employment prospects are provided in Figure 3.

With respect to gender, the income penalty of workers entering start-ups rather than incumbents is slightly larger for men than for women. In year zero, for instance, the difference amounts to  $\in$  3,400 for men and to  $\in$  2,000 for women, which corresponds to percentage income gaps of 18 and 16 percent, respectively. The development of the income gap over the subsequent ten years is remarkably similar for both sexes. Focusing on subgroups defined by age, the youngest workers are experiencing the smallest (but still significant) drawbacks from joining a new establishment, as differences to the balanced control group amount to approximately  $\notin$  2,000 in all years of observation. The income difference increases with workers' age group, both in absolute and relative terms, indicating that the decision to enter a start-up is most harmful for old workers.

Analyzing the development of yearly income for workers of different qualification, we find that the difference to the control group is the largest for workers with a university degree, who earn almost  $\in$  6,000 less even ten years after entry.<sup>18</sup> A similar pattern emerges when we investigate income trajectories for different percentiles of the income distribution. Here, instead of estimating OLS regressions, we estimate

<sup>&</sup>lt;sup>18</sup> The percentage income gap, however, is largest for medium-qualified workers in most periods.

unconditional quantile regressions using recentered influence functions (RIFs) as proposed by Firpo et al. (2009). We focus on the 20<sup>th</sup>, 50<sup>th</sup>, and 80<sup>th</sup> percentile to study the impact of joining a start-up on low-income and high-income earners as well as on the median.<sup>19</sup> Results show that entering a newly founded establishment decreases income most for workers at the 80<sup>th</sup> percentile of the distribution, while the 20<sup>th</sup> percentile is affected to a much smaller extent, indicating that especially workers with high income suffer severe drawbacks (in absolute terms) from joining a start-up as opposed to an incumbent.<sup>20</sup>

Finally, we also test whether the consequences of entering a start-up vary for workers with different previous employment statuses and find that the difference to the control group is slightly larger for workers who performed a job-to-job transition compared to those who came from non-employment. The percentage income gaps, in contrast, are somewhat larger for workers coming from non-employment due to their overall lower income levels. Nevertheless, developments over time are very similar for both groups. Concluding, even though the disadvantages from entering a start-up as opposed to an incumbent are most pronounced for men, old and highly-qualified employees, as well as for workers in the upper part of the income distribution, we find that all subgroups earn lower incomes when joining start-ups rather than incumbents over the whole period of observation.

# 3.3 Results for different subsequent career paths

To explore potential explanations for the significant and long-lasting difference in performance between workers entering start-ups and those joining incumbents, we also investigate income trajectories for workers with different specific subsequent career paths. Specifically, we have a closer look at workers who stay with their initial employer to check whether worse labor market prospects in start-ups occur (only) due to their lower employment stability compared to incumbents. Moreover, to shed light on the relevance of the high failure risk of start-ups for workers' labor market performance, we compare workers who join a start-up that turns out to be successful and does not close down in the early years of business with those entering businesses

<sup>&</sup>lt;sup>19</sup> We cannot analyze higher percentiles of the income distribution, since information on earnings is censored at the contribution limit for social security contributions, which affects approximately 18 percent of observations in our sample. Note that as more high-wage workers are found among those entering incumbents, this should, if anything, lead to an underestimation of the difference between the two groups of workers in our main analysis.

<sup>&</sup>lt;sup>20</sup> In relative terms, income gaps are largest at the median at the beginning of the observation period and at the 20<sup>th</sup> percentile in later years.

that subsequently fail. We further examine the role of start-ups as 'stepping stones' to other positions in workers' subsequent careers. Figure 4 shows the results of these analyses for yearly income.

Focusing on continuing matches, we include only those workers who are still employed with the same establishment which they entered in year zero. Comparing income trajectories of stayers in start-ups with those of stayers in incumbent establishments after balancing (Figure 4a), we see that the difference between the two groups is even more pronounced than in our main analysis, and the gap widens continuously over the observation period. This result indicates that the lower employment stability in start-ups cannot be the only reason for the differences in labor market performance described above. We also find no indication that those workers who remain employed in a start-up over a longer period of time are experiencing steep careers and better earnings prospects (e.g., due to flat hierarchies in newly founded establishments). Instead, continued employment in incumbents seems to lead to steeper wage increases, e.g., due to backloaded compensation schemes or better opportunities for career advancement in internal labor markets.<sup>21</sup>

Additionally, we analyze whether the main reason why workers entering start-ups are less successful on the labor market can be found in the bad economic performance of these establishments. Since many start-ups are failing in their very first years of business (see, e.g., Fackler et al. 2013, Fritsch and Weyh 2006, Mueller and Stegmaier 2015), workers will oftentimes be forced to search for a new job or – in the worst case – become unemployed. Therefore, we divide the group of workers entering a new establishment into those whose employer surviv es over a considerable period of time, i.e., at least five or ten years, and those whose employer closes down within the respective time frame.<sup>22</sup> Figure 4b shows the income trajectories for these specific groups of workers after entropy balancing, indicating that indeed individuals who enter

<sup>&</sup>lt;sup>21</sup> The widening of the income gap between the two groups might also be a result of selective attrition over time, if, for instance, only especially capable workers stay with incumbents, while only the least qualified continue to work for a start-up. To check whether such selection processes drive our results, we compare the characteristics of individuals who remain with their initial employer over time and find no indication for differences in selection dynamics over time for workers in start-ups and incumbents. Additionally, we employ entropy balancing for each period of time separately, to ensure that the two groups are balanced in each year, and then rerun our OLS estimation with these time-varying weights. Our insights remain unchanged.

<sup>&</sup>lt;sup>22</sup> To define establishment exit, we use information on worker flows provided by Hethey-Maier and Schmieder (2013). Instead of categorizing all establishment identifiers which disappear from the data as closures, this allows us to exclude establishments which continue to operate under a different identifier due to mere administrative changes of the identification number or due to take-overs and restructurings, as well as a small fraction which is defined as "unclear" by Hethey-Maier and Schmieder (2013).

a start-up that survives for at least five or ten years, respectively, are performing significantly better than those who joined a start-up which closes down within that time window. Therefore, we also compare the performance of workers entering a surviving start-up with those who initially entered an incumbent establishment, as shown in Figure 4c. However, our results imply that the gap in income between these two groups still amounts to approximately  $\in$  2,000 to  $\in$  3,000 in all periods.<sup>23</sup> Hence, the difference between workers entering start-ups and incumbents cannot solely be explained by the high failure rate of risky new businesses.

Finally, at least one successful strategy for workers joining a start-up might be to use this establishment as a stepping stone to other, potentially more stable or better-payed positions. We define workers using the start-up as a stepping stone as those who leave it reasonably early, i.e., within the first five years after entry, and without an imminent threat of firm exit, i.e., at least two years before closure. Moreover, they are required to take up a job at a different establishment within a maximum of 90 days.<sup>24</sup> We then compare these workers who use the start-up as a stepping stone to a balanced sample of all other individuals entering a newly founded establishment, as presented in Figure 4d, and find that this indeed seems to be a successful strategy. Workers who quickly leave start-ups for positions in other establishments earn approximately € 3,000 more income than the comparison group from year one onwards. Nonetheless, when we investigate the difference in yearly income to all workers who instead joined an incumbent in year zero, we find that the latter are still performing significantly better (see Figure 4e).<sup>25</sup> Thus, even though our analysis of different subsequent career paths after entering a start-up highlights important heterogeneities, it does not reveal any potential channel or strategy through which workers joining a start-up can catch up with, or become even more successful, than workers entering an incumbent establishment.

<sup>&</sup>lt;sup>23</sup> For Britain, Adrjan (2018) also finds that successful start-ups pay higher wages than unsuccessful young firms. Although his results, in contrast to ours, indicate that workers joining successful start-ups initially earn higher wages than those joining an average incumbent, he finds these differences to disappear over time, with disadvantages arising for workers entering successful start-ups instead of incumbents after five years.

<sup>&</sup>lt;sup>24</sup> Since we do not observe the reason for changing an employer in our data, allowing only for a short period of frictional unemployment makes it more likely that our definition of stepping stones just captures workers who leave the start-up voluntarily for a more preferable position.

<sup>&</sup>lt;sup>25</sup> Note that Sorenson et al. (2018) find that using a start-up as a stepping stone might be a successful strategy to catch up with individuals entering an incumbent, if workers are able to subsequently join a large and old establishment. However, their results also indicate that this scenario is highly unlikely due to path dependency in workers' employment relations.

## 4. Conclusions

Although the role of start-ups as employers is often discussed politically and many scholars have analyzed the quantity of jobs created by newly founded firms, the implications of joining a start-up for the individual worker have not been analyzed in depth so far. Therefore, we explore the advantages and disadvantages of entering start-ups instead of incumbent firms, both in terms of remuneration and employment prospects, and investigate whether differences in labor market performance are long-lasting over a worker's subsequent career path. We apply entropy balancing to make both groups of entrants comparable and follow individuals in start-ups and incumbent firms over ten years. Our results imply that workers joining a start-up experience significantly lower income and daily wages, as well as less days in employment and more days of benefit receipt, than similar workers joining an incumbent. These severe drawbacks are persistent over the subsequent ten years after entering the respective establishment and they hold for all groups of workers and types of start-ups analyzed.

Concerning earnings differences between workers entering start-ups and those joining incumbents, our findings are in line with other current research on by Adrjan (2018) for Great Britain and Sorenson et al. (2018) for Denmark, as both studies find long-run pecuniary disadvantages from entering a newly founded firm.<sup>26</sup> We go beyond existing research by showing that persistent drawbacks from joining a start-up can also be found in terms of (un)employment prospects. We also provide insights concerning the role of the higher failure risk and the lower employment stability in start-ups as potential explanations for the observed differences in labor market performance (see also Schnabel et al. 2011). Analyzing workers who remain employed with their initial employer and workers who enter successful vis-à-vis failing start-ups, we still find substantial drawbacks compared to similar workers entering incumbents. When focusing on workers who use start-up employment as a stepping stone to positions in other establishments, we find that even this strategy does not render workers joining newly founded firms as successful as those entering incumbents.

While our main insights imply long-lasting negative consequences from working at a start-up, some limitations of our analysis must be taken into account when interpreting our results. First and foremost, the various indicators of labor market success investigated in this study do not represent all dimensions of job quality. In particular,

<sup>&</sup>lt;sup>26</sup> The findings by Adrjan (2018) slightly differ from our main insights, since he finds workers to earn higher wages in start-ups initially, while long-run disadvantages arise due to higher subsequent wage growth in incumbents.

our data do not allow us to draw any conclusions concerning job satisfaction. Hence, it is possible that workers in start-ups experience especially high levels of job satisfaction due to, e.g., flatter hierarchies or more autonomy (Sorenson et al. 2018). Focusing on remuneration, one must bear in mind that our data do not include information on non-standard means of financial compensation, such as fringe benefits or firm shares. We argue that this shortcoming should not affect our insights, since fringe benefits do not play an important role at the German labor market due to the scope of social security provision by the state (Schmieder 2013), and employee share ownership is not very common in Germany and rarely found in small establishments (Bellmann and Möller 2016). Moreover, the risky nature of start-ups makes it unlikely that firm shares are regarded as an adequate form of compensation by employees. Another limitation could be that our data do not contain self-employed individuals. We thus cannot observe if some workers who were initially employed at start-ups become entrepreneurs themselves, another potential career path that we are not able to analyze. A final, small caveat when interpreting our results is that we do not observe workers' complete employment biographies after entering the respective establishment. However, we claim that the time span of ten years is long enough to observe whether a convergence process sets in and therefore suffices to make meaningful statements on the long-run effects of entering a start-up.

Since all our insights point towards significant disadvantages from entering a start-up, the question arises why workers decide to join newly founded firms at all. One reason might be that individuals are just not well informed about the negative consequences of working for a start-up. Although the high likelihood of failure among new firms is a stylized fact that is often discussed both politically and scientifically (e.g., Geroski 1995, Haltiwanger et al. 2013, Fackler et al. 2013), workers might not be aware of the disadvantages arising even if their employers do not fail. A second potential explanation for workers' decision to enter a new firm could be the different type of work environment. As already mentioned, employment in start-ups is often associated with flat hierarchies, a broader set of tasks assigned to a job, and more responsibility for the individual worker. These factors might compensate workers with strong preferences for such non-monetary job attributes for foregone earnings and worse employment prospects.<sup>27</sup> Finally, it must be noted that newly founded firms often offer opportunities for workers who face disadvantages at the labor market due to, e.g., their

<sup>&</sup>lt;sup>27</sup> An analysis of R&D employees in the U.S. by Sauermann (2018) indeed shows that individuals working in start-ups have strong preferences for job attributes such as autonomy and responsibility but place less importance on job security and income.

age, foreign nationality, or previous unemployment experience (Fackler et al. 2019, Nyström 2012, Coad et al. 2017). Put differently, for some groups of workers, the superior alternative of joining an incumbent may simply not be available. From this perspective, working at a start-up can still offer an opportunity for disadvantaged workers who would otherwise be unemployed, especially if they enter start-ups that prosper and survive or if they intend to use the start-up as a stepping stone for (better) positions in other establishments.

Concluding, since our insights indicate that jobs created by start-ups do not provide workers with the same opportunities for long-run career advancement as those created by incumbents, the role of new firms as job creators should be interpreted cautiously. Even from the viewpoint of individual employees taken in our analysis, the skepticism toward start-up subsidization expressed by some authors (e.g., Santarelli and Vivarelli 2007, Shane 2009) is understandable. In other words, it seems questionable whether start-ups' contribution to employment is substantial enough to warrant the strong political attention and financial support they receive in many countries.

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Table	1:	Selected	characteristics	of new	and incumbent	establishments	(means)
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Variable	New Establishment	Incumbent
Number of employees	2.64	31.20 ***
	(2.67)	(293.61)
Secondary sector (%)	20.35	32.63 ***
	(40.26)	(46.89)
Tertiary Sector (%)	79.65	67.37 ***
	(40.26)	(46.89)
Share of women (%)	47.64	50.60 ***
	(43.52)	(33.40)
Share of full-time workers (%)	68.89	60.50 ***
	(38.14)	(30.64)
Share of marginally employed workers (%)	15.43	22.66 ***
	(25.31)	(26.62)
Number of observations	53,666	126,998

*Sources:* Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Standard deviations are reported in parentheses. \*, \*\*, \*\*\* indicate statistically significant differences between the two groups at the 10, 5, and 1% level, respectively.

Table 2: Selected characteristics of workers entering either a new or an incumbent establishment (means)

Variable	New Establishment	Incumbent		
Women (%)	47.66	46.59	***	
	(49.95)	(49.88)		
Age: 18-24 years (%)	19.21	22.84	***	
	(39.39)	(41.98)		
Age: 25-34 years (%)	35.23	36.57	***	
	(47.77)	(48.16)		
Age: 35-50 years (%)	45.56	40.59	***	
	(49.80)	(49.11)		
Low-qualified (%)	14.88	15.87	***	
	(35.59)	(36.54)		
Medium-qualified (%)	77.85	72.91	***	
	(41.52)	(44.44)		
High-qualified (%)	7.27	11.22	***	
	(25.96)	(31.56)		
Foreign nationality (%)	12.61	11.33	***	
	(33.20)	(31.70)		
Transition from employment (%)	55.89	56.96	***	
	(49.65)	(49.51)		
Part-time (%)	28.62	29.06	***	
	(45.20)	(45.40)		
Years of work experience	5.20	5.21		
	(2.84)	(2.93)		
Years of benefit receipt	1.15	0.82	***	
	(1.69)	(1.45)		
Number of previous employers	4.58	3.97	***	
	(3.84)	(3.60)		
Number of observations	110,201	614,838		

Sources: Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Only individuals of age 18-50 are comprised, excluding apprentices. Years of work experience and years of benefit receipt are truncated at 8 years. Standard deviations are reported in parentheses. \*, \*\*, \*\*\* indicate statistically significant differences between the two groups at the 10, 5, and 1% level, respectively.

Figure 1: Labor market trajectories of workers entering new and incumbent establishments, before and after entropy balancing



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices.



Figure 2: OLS estimates of differences in labor market trajectories between workers entering new and incumbent establishments, before and after entropy balancing

*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS estimates of differences in labor market trajectories as listed in Tables A1 and A2, the grey dashed lines indicate the 95% confidence intervals.

 Figure 3: Subgroup regression estimates (OLS and RIF) of differences in yearly income between workers entering new and incumbent establishments after entropy balancing

 a) By Gender (OLS)
 b) By Age (OLS)



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS (and Rif-Reg) estimates of differences in labor market trajectories, the grey dashed lines indicate the 95% confidence intervals.

Figure 4: OLS estimates of differences in yearly income for workers with different subsequent career paths after entropy balancing



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS estimates of differences in labor market trajectories, the grey dashed lines indicate the 95% confidence intervals.

	a) Yearly Income		b) Average Daily Full-Time		c) Days in Employment		d) Days of Benefit Receipt	
Variable			Earni	ings				
Year 0	18,303.73	(21.43) ***	79.40	(0.06) ***	264.69	(0.14) ***	37.63	(0.11) ***
Year 1	21,978.38	(23.12) ***	83.63	(0.06) ***	308.11	(0.13) ***	29.69	(0.10) ***
Year 2	22,138.95	(24.87) ***	86.29	(0.07) ***	295.00	(0.16) ***	42.90	(0.13) ***
Year 3	22,357.43	(25.95) ***	87.77	(0.08) ***	291.94	(0.16) ***	48.30	(0.14) ***
Year 4	22,720.79	(26.58) ***	89.35	(0.08) ***	292.26	(0.17) ***	50.52	(0.15) ***
Year 5	23,056.27	(26.85) ***	90.53	(0.08) ***	293.62	(0.17) ***	52.34	(0.15) ***
Year 6	23,463.79	(27.09) ***	91.69	(0.08) ***	296.27	(0.17) ***	53.38	(0.16) ***
Year 7	23,947.89	(27.24) ***	92.66	(0.08) ***	299.75	(0.17) ***	51.55	(0.16) ***
Year 8	24,389.72	(27.42) ***	94.17	(0.08) ***	302.51	(0.16) ***	49.65	(0.16) ***
Year 9	24,767.74	(27.68) ***	95.98	(0.08) ***	304.06	(0.16) ***	48.68	(0.15) ***
Year 10	25,240.97	(28.01) ***	98.09	(0.08) ***	306.10	(0.16) ***	47.37	(0.15) ***
Start-Up*Year 0	-5,454.52	(43.38) ***	-21.64	(0.13) ***	-10.88	(0.36) ***	19.78	(0.32) ***
Start-Up*Year 1	-7,076.54	(46.23) ***	-24.34	(0.14) ***	-16.86	(0.35) ***	16.90	(0.30) ***
Start-Up*Year 2	-7,982.83	(50.29) ***	-24.34	(0.16) ***	-30.05	(0.44) ***	26.32	(0.39) ***
Start-Up*Year 3	-8,201.48	(52.80) ***	-24.40	(0.17) ***	-31.68	(0.47) ***	28.49	(0.42) ***
Start-Up*Year 4	-8,320.96	(54.50) ***	-24.85	(0.18) ***	-30.93	(0.48) ***	29.19	(0.44) ***
Start-Up*Year 5	-8,239.11	(55.64) ***	-24.62	(0.18) ***	-29.12	(0.48) ***	28.87	(0.45) ***
Start-Up*Year 6	-8,193.62	(56.75) ***	-24.38	(0.18) ***	-28.18	(0.49) ***	29.72	(0.47) ***
Start-Up*Year 7	-8,093.04	(57.80) ***	-24.08	(0.18) ***	-25.98	(0.48) ***	29.54	(0.48) ***
Start-Up*Year 8	-7,957.13	(58.80) ***	-23.55	(0.19) ***	-23.19	(0.47) ***	28.22	(0.47) ***
Start-Up*Year 9	-7,885.24	(59.81) ***	-23.09	(0.19) ***	-21.39	(0.47) ***	26.94	(0.47) ***
Start-Up*Year 10	-7,905.55	(60.84) ***	-22.72	(0.19) ***	-20.87	(0.47) ***	26.33	(0.47) ***
Number of obs.	7,242,162		5,030,629		7,242,162		7,242,162	
R²	0.5803		0.7861		0.8547		0.1865	

Table A1: Results of OLS regression of various labor market outcomes in the unbalanced sample

Sources: Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. Notes: Ordinary least squares regressions. Standard errors (reported in parentheses) are clustered by individual workers. \*, \*\*, \*\*\* indicate statistical significance at the 10, 5, and 1% level, respectively.

	a) Yearly Income		b) Average Daily Full-Time		c) Days in Employment		d) Days of Benefit Receipt	
Variable			Earn	ings				
Year 0	15,804.24	(25.59) ***	68.93	(0.08) ***	259.79	(0.20) ***	52.03	(0.19) ***
Year 1	18,688.50	(27.44) ***	71.99	(0.08) ***	301.06	(0.19) ***	41.44	(0.18) ***
Year 2	18,418.55	(29.75) ***	74.67	(0.09) ***	283.00	(0.24) ***	57.40	(0.22) ***
Year 3	18,417.96	(31.04) ***	75.94	(0.10) ***	278.37	(0.26) ***	64.13	(0.24) ***
Year 4	18,620.72	(31.85) ***	77.10	(0.10) ***	278.41	(0.26) ***	66.88	(0.25) ***
Year 5	18,879.33	(32.34) ***	78.15	(0.10) ***	279.33	(0.27) ***	68.91	(0.26) ***
Year 6	19,269.10	(32.81) ***	79.37	(0.10) ***	282.40	(0.27) ***	70.31	(0.27) ***
Year 7	19,765.24	(33.21) ***	80.31	(0.10) ***	286.92	(0.26) ***	68.02	(0.27) ***
Year 8	20,240.61	(33.52) ***	81.89	(0.10) ***	290.75	(0.26) ***	65.30	(0.27) ***
Year 9	20,609.86	(33.97) ***	83.82	(0.11) ***	292.62	(0.26) ***	63.70	(0.27) ***
Year 10	21,030.33	(34.44) ***	85.93	(0.11) ***	294.83	(0.26) ***	61.94	(0.27) ***
Start-Up*Year 0	-2,955.03	(45.56) ***	-11.18	(0.14) ***	-5.98	(0.39) ***	5.38	(0.36) ***
Start-Up*Year 1	-3,786.66	(48.51) ***	-12.71	(0.15) ***	-9.81	(0.38) ***	5.15	(0.34) ***
Start-Up*Year 2	-4,262.43	(52.84) ***	-12.72	(0.17) ***	-18.05	(0.48) ***	11.82	(0.43) ***
Start-Up*Year 3	-4,262.00	(55.45) ***	-12.57	(0.18) ***	-18.11	(0.51) ***	12.66	(0.47) ***
Start-Up*Year 4	-4,220.89	(57.22) ***	-12.60	(0.19) ***	-17.09	(0.52) ***	12.82	(0.49) ***
Start-Up*Year 5	-4,062.16	(58.45) ***	-12.24	(0.19) ***	-14.84	(0.53) ***	12.29	(0.50) ***
Start-Up*Year 6	-3,998.93	(59.65) ***	-12.07	(0.19) ***	-14.32	(0.53) ***	12.78	(0.52) ***
Start-Up*Year 7	-3,910.39	(60.80) ***	-11.73	(0.20) ***	-13.15	(0.52) ***	13.07	(0.52) ***
Start-Up*Year 8	-3,808.02	(61.84) ***	-11.27	(0.20) ***	-11.43	(0.52) ***	12.57	(0.52) ***
Start-Up*Year 9	-3,727.36	(62.91) ***	-10.92	(0.20) ***	-9.96	(0.51) ***	11.92	(0.52) ***
Start-Up*Year 10	-3,694.91	(64.01) ***	-10.56	(0.21) ***	-9.59	(0.51) ***	11.75	(0.52) ***
Number of obs.	7,242,162		5,030,629		7,242,162		7,242,162	
R <sup>2</sup>	0.5381		0.7587		0.8247		0.2388	

Table A2: Results of OLS regression of various labor market outcomes in the balanced sample

Sources: Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. Notes: Ordinary least squares regressions. Standard errors (reported in parentheses) are clustered by individual workers. \*, \*\*, \*\*\* indicate statistical significance at the 10, 5, and 1% level, respectively.

Figure A1: OLS estimates of differences in labor market trajectories between workers entering new and incumbent establishments, including lags of labor market outcomes in the balancing procedure



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS estimates of differences in labor market trajectories between workers entering a start-up or an incumbent, the grey dashed lines indicate the 95% confidence intervals.

Figure A2: OLS estimates of differences in labor market trajectories between workers entering new and incumbent establishments, sample restricted to establishments with not more than 20 employees in year zero.



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments with not more than 20 employees in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS estimates of differences in labor market trajectories between workers entering a start-up or an incumbent, the grey dashed lines indicate the 95% confidence intervals.

Figure A3: OLS estimates of differences in labor market trajectories between workers entering new and incumbent establishments, sample restricted to establishments in research intensive manufacturing sectors



*Sources:* Integrated Employment Biographies (IEB); Establishment History Panel (BHP); authors' calculations. *Notes:* Sample includes only workers entering establishments in research intensive manufacturing sectors in the years 2000-2004, excluding agriculture, energy and mining, and the public and nonprofit sectors. Sample comprises individuals of age 18-50, excluding apprentices. Graphs show the OLS estimates of differences in labor market trajectories between workers entering a start-up or an incumbent, the grey dashed lines indicate the 95% confidence intervals.