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Training, Soft Skills and Productivity: Evidence from a Field Experiment in Retail

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Training, Soft Skills and Productivity: Evidence from a Field Experiment in Retail*

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May 3, 2019

Abstract

Understanding the causal association between skills and productivity is essential for designing effective training programs. This paper evaluates an intervention aimed at boosting leadership and communication skills among store managers and sales associates from a large Latin American retailer. The empirical analysis is carried out using longitudinal information gathered by the firm and through two skills surveys. The identification exploits the experimental design in the context of a difference-in-difference strategy. The results indicate large positive effects of the training program on store-level productivity. We further link these findings to individual-level performance measures. In particular, we document positive effects on total sales and numbers of transactions for all workers. Regarding the mechanisms, we provide evidence suggesting that the intervention was more effective in boosting leadership skills than communication skills. Spillovers from trained managers to untrained sales representatives also contribute to the main effects. Our findings point towards the possibility of increasing productivity through training programs targeting critical skills.

Keywords: Socio-emotional skills, training, productivity, experiments with firms.

JEL classification: C93, J24, M53, O15.

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1 Introduction

The study of the interrelation between human capital accumulation, training and productivity has a long-standing tradition in economics (Becker, 1964). Among workers, on-the-job training is the building block of the human capital accumulation process. In fact, almost half of the human capital individuals accumulate in their lifetime is associated with investments and activities related to work (Heckman, 1998). However, it was not until recently that the economic literature focused on the role of skills as key underlying drivers of the human capital accumulation process. During the last decade, a growing body of research has documented the essential role of socio-emotional skills - commonly referred to as “soft” skills - as determinants of labor market productivity, both at the individual and aggregate levels (Heckman and Kautz, 2012, Daruich, 2018). These include dimensions such as communication, teamwork, and planning and organizing, among others.¹ This has triggered new interest in the effects of on-the-job training programs as, in principle, they impact labor market outcomes through skill development (Bassanini et al., 2005, Barrett and O’Connell, 2001, Dearden et al., 2006, Konings and Vanormelingen, 2015).

In this paper we use a field experiment to estimate the impact of a training intervention intended to modify critical socio-emotional skills – leadership and communication – among workers of one of the largest retailers in Latin America. Through an interactive five-day program, managers from a randomly selected sample of stores participated in activities specifically designed to improve their skills to succeed as team leaders. The program encourages the managers to think about their capabilities and provides techniques to develop new management skills. Likewise, through a two-day program, selected sales associates were trained to develop critical communication skills. The technical content of the sessions includes themes such as complaint handling, advising customers, and understanding the importance of customer service as core to success.

To assess the impact of the intervention, we exploit administrative information provided by the firm, and two skills surveys (baseline and follow up). To secure its identification, we complement the experimental design with a difference-in-difference approach. We quantify the

¹Skills have been shown to determine individual-level labor market outcomes, with direct repercussions on productivity, economic growth and other economic dimension, in addition to the well-established effects of skills on economic growth (OECD, 1994, 2015; Hanushek and Woessmann (2008, 2012), among others), the evidence has documented the impact of skills on learning and multiple academic outcomes, the probability of getting and keeping a job, wages, occupational choices and also a series of social and health behaviors. For example, Heckman et al. (2006); OCDE, 2015; Hanushek and Woessmann (2008, 2012); Urzúa (2008); Prada and Urzúa(2017); Almlund et al. (2011); Heckman and Kautz (2012); Heckman and Kautz (2014a,b); Borghans et al. (2008). For further details on this literature see Roberts et al. (2007) or Borghans et al. (2008).

effect of the training on direct measures of productivity both at the store and individual level. Moreover, to shed light on the economic mechanisms, we investigate the extent to which our findings are explained by changes in leadership and communication skills.

Our main results can be summarized as follows. We document positive and significant effects of the training intervention on store-level productivity, as well as on individual labor productivity, measured by individual monthly sales and number of transactions. We also find evidence supporting the hypothesis that the training boosted leadership and management skills among store managers. However, we could not uncover any impact of the training on measured communication skills among salespersons.

The contribution of our analysis to the literature is three-fold. First, given the experimental design, we use the exogenous variation in training participation to identify the causal effect of training in the context of a private firm. We also shed light on to whom the training should be directed to within a company (managers, sales associates or both groups), and the differential impact associated with that decision. Second, rather than analyzing a wide spectrum of skills, we provide evidence of the impact of a program specifically aimed at boosting two socio-emotional dimensions. This allows us to narrow down the mechanisms explaining the estimated effects of the intervention. Third, by using detailed information provided by the firm, we can assess the causal impact of training on direct productivity measures (sales and transactions). Despite the long-standing literature analyzing the impact of on-the-job training, to the best of our knowledge, ours is the first paper providing direct evidence on these dimensions.

Like any other experimental study, our findings should not be considered “externally valid”, that is, they cannot be directly extended to other settings (Deaton and Cartwright, 2018). Nonetheless, the analysis provides useful insights for policy making. This as our findings not only show how a tailored made on-the-job program can increase the stock of skills during adulthood but also confirm it is possible to improve labor market outcomes by targeting malleable skills. In addition, we highlight the importance of public-private collaborations for policy design. The identification of critical skill gaps for this project would have been impossible without a direct and fruitful collaboration between the research team and the private firm. Thus, beyond our main results, this study also throws lessons about the characteristics of effective training programs. These might be particularly relevant for developing countries, where labor productivity growth is sluggish and significant public resources are allocated to extemporaneous training programs. In

sum, our paper speaks to the decades of public efforts seeking to implement effective initiatives with direct impacts on labor market outcomes

Thus, in addition to the immediate results of our experiment, the study also provides insight about the characteristics of effective training programs.

The paper is structured as follows. Section 2 summarizes the literature review. Section 3 describes the intervention. Section 4 describes the experimental design. Section 5 presents the data sources used and exploratory results. Section 6 presents the store-level analysis and main results, while section 7 provides the individual-level analysis. Finally, section 8 provides conclusions and final reflections.

2 Literature Review

The empirical identification of the causal impact of skills training programs on productivity represents an industrious and challenging task (Heckman et al., 1999). First, the non-random selection of workers into the training prevents the direct estimation of causal effects, as these cannot be unraveled from other factors that might also be associated with both individual productivity and participation in the training program. Second, constructing good proxies for workers' productivity is rather difficult. While wages are often used to estimate returns to training, these may not capture the overall impact of participation on productivity (Dearden et al., 2006).² Third, training programs are heterogeneous: they vary greatly in terms of quality, type, purpose, intensity and objective. As a result, to a large extent, the existing evidence on the impact of training is mixed (Kluve, 2016, Urzúa and Puentes, 2010, Mitnik et al., 2016, Bassanini et al., 2005, Ibarrarán et al., 2015). This prevents the identification of effective mechanisms to increase worker's labor market productivity, which in turn leads to inadequate guidance for policy-making.

Our paper contributes to the literature in the three aforementioned areas. It uses exogenous variation in training participation to identify its causal impact. In addition, it exploits rich longitudinal data to construct direct productivity measures. And, since the intervention was designed to modify two specific skills, it narrows down the main mechanisms behind any estimated effect.

²Conti (2005), Dearden et al. (2006), Almeida and Carneiro (2009), Konings and Vanormelingen (2015) use individual-level productivity measures as the outcomes of interest and Barrett and O'Connell (2001) examines sales. See González-Velosa et al. (2016) for estimates in the context of Latin American countries.

Despite the long-standing literature on the impact of training programs, only a handful of studies are closely related to ours. De Grip and Sauermann (2011) and Adhvaryu et al. (2018) quantify the impact of on-the-job training on direct measures of productivity, using an experimental design. Both papers do this in the context of a private firm. Specifically, De Grip and Sauermann (2011) assess a task-specific intervention combining formal coaching addressing tips to improve productivity and learning by doing. They find that participation in a 38-hour program leads to a 9 percent increase in productivity. The paper also presents evidence of externalities from treated workers on their untreated peers. In particular, an increase of 10 percentage points in the share of treated peers leads to a productivity increase of 0.45 percent.³ By focusing on a training program targeting two socio-emotional skills, which can be conceptualized as more general in nature than a task-specific intervention, our analysis complements the results from this study.

More recently, Adhvaryu et al. (2018) evaluate the impact of soft skills training on workplace outcomes among female garment workers in India.⁴ The training targeted a wide range of skills including communication, problem solving, hygiene, reproductive health, among others.⁵ Based on an experimental design, the authors report a 12 percent increase in productivity, a 0.5 percent increase in wages after program completion, and a large net return to on-the-job soft skills training (250 percent 9 months after program completion). Our results are in line with these findings. However, since we mainly focus on only two dimensions of soft skills, we can narrow down the mechanisms explaining the estimated effects of training. In fact, neither De Grip and Sauermann (2011) nor Adhvaryu et al. (2018) document the direct impact of training on the targeted skill dimensions.⁶

Our paper also speaks to the strand of the literature analyzing spillovers arising from the

³The experiment analyzed was implemented in an in-house call center of a multinational mobile network operator in the Netherlands. The training consisted of a discussion with a coach on the skills needed to improve productivity -i.e. reduce the time an agent needs to handle a customer call- how these skills could be improved and practical tips. The training combined formal conversations with learning by doing.

⁴The paper focuses on the number of pieces (garments) produced and the efficiency of the production process. Efficiency is calculated as the number of pieces produced divided by the target quantity of pieces per unit of time and, alternatively, as the Standard Allowable Minute (SAM). SAM is defined as the number of minutes that should be required for a single garment of a particular style to be produced.

⁵The paper investigates the Personal Advancement and Career Enhancement (P.A.C.E.) program. It was designed and first implemented by GAP Inc. The program consists in a 80-hour training focused on improving life skills such as time management, effective communication, problem-solving, and financial literacy for its trainees. It also included additional modules on Water, Sanitation and Hygiene and General and Reproductive Health.

⁶Adhvaryu et al. (2018) only use survey responses to support the hypothesis that the program increased the stock of soft skills in the month after program completion.

human capital accumulation process. The experimental design in Adhvaryu et al. (2018) allows for spillovers to untreated workers within the five garment factories in Bengaluru (India), both through the transfer of skills as well as through production complementarities. Likewise, De Grip and Sauermann (2011) reports experimental evidence for general externalities from treated workers on their untreated teammates. However, these studies do not analyze spillovers across job titles (e.g., from managers to sales associates and vice-versa). In this way, we shed light on to whom the training should be directed to within a company (supervisors, employees or both), and estimate the differential impacts from that decision. Thus, our findings are related to the recent evidence on the effectiveness of management practices (Gosnell et al., 2019).

Finally, our paper connects to the research highlighting the importance of examining skills at a more granular level (Kern et al., 2013). This in turn allows interrelationships across dimensions to be uncovered (Borghans et al., 2008, Cunha et al., 2006). For instance, using data from the United States, Prada and Urzúa (2017) document how manual and mechanical skills, which can be conceptualized as cognitive dimensions, have distinct effects on labor market outcomes relative to those previously documented using more conventional facets. This also links to the recent findings on the role of routine and non-routine tasks in changing labor markets (Acemoglu and Autor, 2011). By investigating the potential complementarity between different soft-skills in the context of a training program we contribute to these previous efforts.

3 The intervention

The training intervention was designed to enhance two critical skills: *leadership* and *communication*. It was offered to the employees of a major retail firm in Latin America, which has a large market share in the footwear, clothing and accessories sector in Chile. Its business consists of importing, advertising and distributing products from international and domestic brands.⁷ The cornerstones of the company are the strength of its brands, the high quality of the products, and the customer service provided to the client.

Firm's sales operations are conducted primarily in large shopping centers (malls) and, more recently, in small outlet malls. By 2014, the firm had 324 stores located throughout the country. Each store has one manager who oversees the operations, coordinates and supervises the work

⁷The company has a large portfolio of trademarks to cover the different segments of the footwear and clothing market. In 2015, it managed 15 consolidated brands.

of the sales associates and other employees and reports directly to the coordinating manager of all the stores in each brand (chain manager). Almost 78% of the firm's employees work in the stores as sales associates.

The design of the training program. In contrast to most of the existing literature, the training program examined herein was designed to address specific skill needs within an organization. To identify these critical dimensions, we first implement an ex-ante assessment of the training needs of the firm and identify major skill gaps among its employees.

The training needs assessment was conducted using multiple activities, including (i) interviews and focus groups with the business executives to assess the future of the business and the skills that employees need to meet the company's expectations; (ii) focus groups with 12 store managers to identify the skills needed to increase productivity within each location, and the methods considered most effective and convenient to address the identified skill gaps; (iii) interviews with managers and sales associates at 14 stores to evaluate how staff perceived the business needs as well as the skills they require; (iv) the collection and analysis of a skill survey designed to capture self-assessed current and future skills needs, and preferred learning methods; and (v) 360-degree reviews with a selected group of 16 managers, their supervisors (chain managers), and their staff (sales associates). These provided an in depth understanding of skill levels within the firm.

From these exercises, three major training needs emerged: (i) coaching and leadership skills for managers, including building team confidence and good teams, dealing with different personalities and learning styles; (ii) selling and communication skills, building confidence in communication, and frustration and conflict management; and (iii) technical training such as brand knowledge and specifications. Thus, after considering the internal knowledge of the production function of the firm, we singled out *communication* and *leadership* as the two critical skills to be targeted by the intervention.

To incorporate the large differences between the type of activities and tasks performed by store managers and sales associates, we designed a training subprogram for each job profile. The training intervention was designed by one of the leading skills and workforce development organizations in the UK.⁸ Each training subprogram included: (i) the design of a curriculum

⁸People 1st is an organization with more than 50 years of experience designing and implementing training strategies for the retail, hospitality, travel, tourism and passenger transport industries.

(training content) aligned with the assessed needs, and (ii) the design of a structured plan to ensure high-quality in the delivery of the training, which included a careful selection of master trainers, extensive training for master trainers to ensure homogeneous training experiences to all employees, and close monitoring of the progress of both the master trainer and the employees receiving the training.

Training store managers. Through an interactive five-day training, managers learned key confidence building skills to succeed as *leaders*, focusing on developing *individual leadership skills* to exploit the human capital potential of sales associates and, to maintain satisfied and productive teams. The program encourages the managers to think about their own management skills and develop new management techniques. An important component of the program was its interactive and dynamic nature, which explains why a large fraction of the training was devoted to practical activities using concrete examples and promoting active participation supplemented with supervised practice of what has been learned. The course was organized in modules offered in 7-hour sessions, for a total of 30 hours of training. Table 1 presents a detailed description of the content of the training each day.

The sales personnel program. Through an interactive two-day training, store employees learned central *communication* skills in the context of their daily tasks and specific technical content in areas such as complaint handling, understanding their product to advise costumers, technical knowledge on specific detail of the products, and understanding the importance of customer service as core to success. Thus, this training module aimed to develop effective communication skills to improve sales strategies, to communicate better with clients and to enhance internal communication with other members of the store. The program encourages collaborative strategies using real-life experiences of good and bad service and promote the understanding of the factors influencing customers choices. The course was organized in twelve 60-minute modules. Table 2 displays a detailed description of the modules.

Quality assurance. As the economic literature shows, high-quality trainers (e.g., teachers) are essential to ensure a high-quality learning experience with long-lasting effects (Chetty et al., 2014). This principle explains the important role of the "Master trainers" program, one of the components of the intervention.

The main objective of the “Master Trainers” program was to train the individuals in charge of delivering both the store manager and team training programs. To this end, a group of senior managers was carefully selected by the firm. Selected managers had to be perceived by their supervisors as suitable persons to train the trainers, particularly recognized for their exceptional leadership qualities and high motivation. Their training consisted of special sessions delivered by a senior Spanish-speaking Master trainer from People1st. Overall, eight employees became fully qualified trainers for the management training and seven for the salespersons training.

Another important component to ensure the quality of the intervention was the full involvement of the firm, with the exception, of course, of any decision or action leading to the selection and assignment of stores to treatment and control groups. This strategy aligned the intervention’s activities with the firm’s capacity to sustain and build on training over time. Given its organizational and learning culture, this helped to guarantee that employers knew why they were being trained for and what the expectations of their supervisors would be.⁹

Dates of training. To avoid affecting the normal operation of the stores, the training for the managers was distributed over a period of 5 weeks (1 day per week), between the fourth week of May and the fourth of June 2016. The training for the sales team took place in two consecutive days during the month of August.

4 Experimental design

Sampling Frame. Our sampling frame used as input the list of all stores reported by the firm as of October 2015. We selected those located in the largest municipalities from the three largest regions of Chile: Viña del Mar and Valparaíso in region V; Talcahuano and Concepcion in region VIII and Santiago in Metropolitan Region-RM. We selected these municipalities for three reasons. First, they contain 60% of the operating stores in the country.¹⁰ Second, overall these three regions represent a large proportion of the country’s economic activity (57% of Chile’s

⁹It is worth mentioning that store managers and sales associates completed learning assessments at the end of each training day. Only those who attended all sessions, completed the practical exercises and the learning assessments, received a certificate as proof of a successful participation. In practice, only the attendance condition was binding, because all the participants with full attendance completed the practical exercises and fulfilled the learning standards.

¹⁰Out of the 324 stores throughout the country, 222 were in the three regions included in the experiment, and 194 in the five largest cities within these regions.

GDP in 2015). Third, given the geographic proximity, the stores in these regions/municipalities offered practical advantages given the intervention's budget.¹¹

In addition, we consider all stores located outside of a shopping mall to be ineligible for inclusion in the experiment. This decision was made after identifying critical differences between units located inside or outside shopping centers (traffic of potential clients, security provisions, hours of operation, total sales, knowledge and merchandise transfer between stores, etc.). This left us with a total of 157 potential stores to be considered for the intervention.

Finally, to secure the existence of a large number of comparable treatment and control units, we removed those stores located in shopping malls with less than three stores. The 157 stores were operating in 34 shopping malls but only 22 shopping malls have 3 or more stores. Thus, after applying the three restrictions on the sample -i.e., geographical location, shopping mall location and shopping mall size- we ended up with 138 stores eligible for the intervention. These stores represent nearly 43% of all the stores nationwide, 76% of the units located in the selected municipalities and 90% of the annual sales in 2014.

Randomization process. The randomization process involved allocating stores into three different categories: the control and two treatment groups. For the stores randomly assigned to the first treatment group (T1), their managers received the training program focused on leadership, management and coaching skills. For the second group (T2), both manager and sales associates received training. While the former received the manager training, sales associates received the training program focused on communications and sales. In the last group (control C), neither managers nor staff were trained.

We exploited the characteristics of the sample and stratified by shopping mall. This means that in each of the 22 malls, we randomly select stores into one of three groups. To maximize the statistical power of the intervention with multiple treatment groups, we set the number of stores in the control group to be the largest of all categories.¹² Table 4 presents the distribution of the selected 138 stores across the three groups -T1, T2 and C-.

Compliance and Identification. Out of the 82 managers in the stores assigned into the initial treatment groups, 71 ended up attending the training sessions. Among sales associates

¹¹See Table 3 for more details.

¹²This decision was made to minimize the sum of the minimum detectable effect for the two treatments. See Bloom (1995) and Duflo et al. (2006) for more detail.

the final number was 176 (out of the original 179). The differences between the initial and actual numbers are explained by the company's normal turnover, licenses and other activities. In addition, given the nature of the business and the impossibility to interfere with firm's strategy decisions, after the initial random assignment few managers moved to different locations. Since the firm was not aware of the initial assignment, these decisions were exogenous to the intervention. However, reallocation and new hires implied that some of the stores originally assigned to the full intervention group (T2) ended up with a new and untrained manager. Instead of discarding these cases, we took advantage of them. We made use of these exogenous events to define a new intervention group, (T3), composed by stores where only sales associates (sellers) received training.

Table 4 describes the final structure of our sample. It compares the number of stores in the initial treatment assignment with the stores in the final treatment groups, while Table 5 compares the number of employees to be trained -managers and sales associates- following the initial treatment assignment with the number of employees that actually received and completed the training.

The research team anticipated that firm's strategic decisions could alter the original structure of the training program. However, given the nature of the intervention, forcing the firm to put on hold the re-allocation of managers and other personnel across stores would have compromised its participation. This is a trade-off any experimental approach dealing with firms must face.

Table 6 explores the potential consequences of these managerial decisions. It presents baseline balance checks for a set of relevant variables and two different time periods. The new treatment group (T3) does not display differences in pre-treatment variables relative to the others as, in general, we fail to reject that mean differences across groups are statistically different from zero.¹³ However, the number of stores in T3 is smaller than those in T1 and T2. This fact may limit our chances to detect small size effects.

In despite the confirmatory results from the store-level balance tests, we recognize imperfect compliance adds additional complexity to our intervention. Therefore, to further shield our identification arguments from threats emerging from firm's strategic behavior, we complement the experimental design with quasi-experimental methods. As described below, given the potential consequences of firm's decisions regarding the pre- and post-treatment assignment of workers

¹³An exception is the dummy variable "Region XIII" (RM). In this case, we find small pre-existing differences in the baseline. Our empirical analysis takes this into account.

across treated and control stores, we exploit the rich store- and individual-level longitudinal information, the timing of the intervention, and the treatment/control status to implement a Difference-in-Difference strategy (DD).

5 Data and Exploratory Results

Our empirical analysis uses three sources of information: store and employee-level records provided by the firm, including daily sales and number of transactions; and two skill surveys (baseline and follow up) characterizing employees' cognitive and socio-emotional abilities as well as their socio-economic background.¹⁴

Firm's administrative records. The firm's detailed records include (i) daily sales from all transactions for the period January 2014 - November 2016 for all 324 locations, including employee's identifiers for 8,553 individuals who worked at the firm at least one day during the period; ii) worker-level monthly earnings from January 2014 to September 2016, including positions, promotions, tenure at the company (and specific location), sales commission and other demographic variables (age and gender); and iii) general information about stores, including geographical location (region and municipality), size (square meters and number of workers) and whether or not it is located at a shopping center. Figure 1 presents total daily store sales and number of daily store transactions for the period 2014-2106. As expected, both sales and transactions follow a clear cyclical pattern, with significant increases on specific dates (e.g., Mother's Day and Christmas). Our empirical strategy takes these patterns into account. Figure 2 illustrates how the longitudinal employee-level data allows us to characterize individuals' progression within the firm. For a given individual, we can identify her job position, earnings, and sales commissions since she was hired by the firm.

Table 7 presents summary statistics for the variables of interest by treatment status between 2014 and 2016. The outcomes include total daily store sales, daily sales per worker, daily sales per transactions, total daily transactions and daily transactions per worker. Average daily sales range between \$1.1 million and \$1.4 million Chilean pesos (CLP) (between 1,800 and 2,300 US dollars) with daily transitions reaching approximately 8 per day per worker. The average

¹⁴The information on sales and transactions is recorded in real time, but we use daily aggregates to reduce the noise created by the natural dynamics of a day in a store.

number of workers per store is 7 across control and treatment groups. The variable “variety” is a store-level indicator of the degree of specialization, while “Brand” and “Brand-Short” identify the main brands sold at each location.¹⁵

Skill Surveys. As discussed in Section 3, the training needs assessment identified opportunities in two areas of socio-emotional abilities: Coaching/leadership skills for managers, and selling/communication skills for the sales force.

We conducted two surveys to characterize the skills of managers and sales associated working for the organization. We use a website specially designed for this purpose. The first round (baseline survey) gathered information for 1,661 workers during the months of November and December of 2015. The second round (follow up) gathered information on 1,569 workers collected over the months of November and December of 2016 with few responses in early January of 2017. Given the high mobility of workers within the firm, we count with information on cognitive and socio-emotional skills for 2,445 individuals. From those, 781 workers reported skills in both surveys.

The battery of socio-emotional tests included communication and leadership skills as well as meta-cognitive strategies (planning activities and critical thinking), self-efficacy (self-perception of ability to organize and achieve goals), grit and the five big characteristics of the “Big Five” personality traits (openness to experience, conscientiousness, extraversion, agreeableness, and neuroticism). Information on cognitive skills included Raven’s Progressive Matrices, reading comprehension, and mathematical knowledge. Additionally, the surveys gather demographic information (age, gender, schooling level, parental education, and other socio-economic characteristics) and variables describing the past experience at the company as well as expectations about the future. Table 8 presents summary statistics of skills as well as other variables of interest.

By combining firm’s records and individual responses to the surveys we can first examine the empirical association between different skill dimensions and individual-level outcomes of interest. In particular, we link leadership and communication to the probability of promotions within the firm (from sales to store management) and the average sales per transaction (December 2015).¹⁶

¹⁵The variable “Variety” records how many product categories are sold at the store, where the categories are: footwear, clothing, accessories and others. “Brand” takes a different value for each of the 15 brands the firm owns. “Brand_short” groups the 15 brands into four categories depending on the average sale price per item.

¹⁶The leadership and communication variables represent standardized versions (mean zero and variance one)

We use the first round of the skill survey to assess the importance of leadership and communication. Figures 3 and 4 display the contour plot of the association between worker's leadership and communication skills and her total sales in 2014 for sales associates and managers, respectively. In general, higher levels of both skills are associated with more sales, confirming the conclusions drawn from the original training needs assessment. However, different patterns appear for workers depending on their roles in the firm. In the case of sales associates, higher sales are present for individuals with high levels of both skills dimensions. However, the combination of high communication and low levels of leadership also correlates with high sales levels. In contrast, for managers the combination of high levels of leadership with relatively low scores in communication are correlated with the largest values of total sales in 2014.

Tables 9 and 10 explore these associations in further detail. The first one presents the results of a probit model of promotions (by December of 2015) on the characteristics of the employees. Regardless of the specification, we estimate that one standard deviation increases in "leadership" at baseline increases the probability of promotion to store manager between 15 and 20 percentage points. This represents a sizable magnitude, as the average probability of promotion within the sample is only 2.6%.¹⁷ Table 10, on the other hand, presents the results of a regression of average sales on individual's characteristics. These findings confirm a significant and positive correlation between communication and sales per transaction. The results are robust to different specifications.

Thus, while the intervention's emphasis on leadership and communication was a direct result of a qualitative analysis, it is interesting to confirm their correlation with productivity (sales) and career progression (promotion).¹⁸

of the averages generated from two modules of the Social and Personal Competencies Scale (CPS for its Spanish acronym "Escala de Competencias Personales y Sociales"). The CPS is a non-cognitive test that measures the effectiveness of life skills in developing positive attitudes and values. The test measures six basic competencies: leadership, empathy and communication skills, behavior in situations of conflict, self-esteem and abilities to relate with others. Our analysis exploits the first two dimensions. The scale was developed to evaluate the life skills training component of a training program in the Dominican Republic. See for example, Ibarrarán et al. (2009), Ibarrarán et al. (2014), Ibarrarán et al. (2015). Refer to Prada and Rucci (2016) for more details on the CPS.

¹⁷It is important to clarify that the results of the skill measurements are confidential and were not shared with the executives of the company. Therefore, promotion decisions could not have been influenced by these tests.

¹⁸To a large extent, promotions within the organization are based on individual productivity measures, e.g., total sales.

6 Store-level analysis

Since the intervention was carried out at the store-level, we first present store-level results. The next section presents our main findings using individual-level data.

As previously stated, our experimental design of the intervention should lead to the direct identification and estimation of its impacts. However, as a result of the natural adjustments made by the firm (e.g., hiring and firing decisions, transfers and store closures), we exploit the rich longitudinal data and the random assignment of treatment to implement a Difference-in-Difference strategy (DD). Formally, we estimate the following regression model:

$$Y_{j,t} = \beta_0 + \beta_1 D_j + \beta_2 B_t + \beta_3 D_j B_t + \beta_4 X_j + \beta_5 g_t + E_{j,t}, \quad (1)$$

where $Y_{j,t}$ represents the outcome of interest of store j (e.g., sales) in period t , D_j is a binary indicator taking a value of one if store j was treated and equals to zero if belonged to the control group, B_t defines the period (before or after the training program), X_j is a vector of store-level controls measured at baseline (e.g., geographical location, type of products, number of employees, etc.), g_t is a set of time controls (year-week fixed-effects and weekend dummies), and $E_{j,t}$ is the error term. The parameter of interest is β_3 .

A central feature of the intervention is the existence of three different treatment groups: stores where only managers were trained ($T1$), both managers and salespeople were trained ($T2$) and only salespeople were trained ($T3$). Therefore, given the nature of the intervention, we focus on the differences between each of these groups relative to the untreated stores. To this end, we estimate separate regressions for each treatment ($T1, T2, T3$) and define D_j accordingly.

With respect to the post-treatment period, for the manager intervention ($T1$) we use the first 147 days after the end of the training, whereas for sales associates ($T2$ and $T3$) we consider the first 77 days of post-treatment data. In addition, given that we only have post-treatment information for the weeks 25-46 of 2016, we estimate equation (1) restricting the sample to the same weeks in the pre-treatment years (2014 and 2015). In this way, we secure the comparability of our outcomes, avoiding the seasonal patterns characterizing the first weeks of the year.¹⁹ We

¹⁹Figure 5 shows the weekly evolution of sales during the years 2014, 2015 and 2016. Mother's Day in May and Christmas season generate changes in the dependent variable that cannot be captured by the post-treatment data. In consequence, we restrict the sample to consider the cyclical behavior of sales in the empirical estimation. Table 11 presents the summary statistics for this sample. Our findings are similar to those obtained without taking into account the cyclical behavior of store sales.

interpret our point estimates as the short-term effects (as of November 16, 2016).

Additionally, for stores in which both the manager and sales associates received training (T2), we exploit the timing of the intervention to investigate potential synergies emerging from the program. More precisely, within those stores, we use the longitudinal data to assess the impact of the training in two different post-treatment periods: after the manager has been trained but the sales team has not, and after all employees (sales associates and manager) have gone through the training program.²⁰

Let B_t^1 be a dummy variable capturing the period in which only the store manager has been treated ($B_t^1 = 1$ if post-treatment, and 0 otherwise). Likewise, let B_t^2 be a dummy variable identifying the period after all store employees have been trained. Thus, we extend equation (1) and estimate the following regression models for the analysis of the impact of T2:

$$Y_{j,t} = \alpha_0 + \alpha_1 D_j + \alpha_2 B_t^k + \alpha_3 D_j B_t^k + \alpha_4 X_j + \alpha_5 g_t + u_{j,t} \text{ with } k = \{1, 2\}, \quad (2)$$

where $u_{j,t}$ is the error term and α_3 is the parameter of interest. In this case, the identification strategy remains the same as in model (1).

Findings. We investigate the impact of the training program on two outcomes of interest: total daily sales and total transactions per day.

In our sample, the average daily sales in stores is approximately 1.1 million CLP (1,630 dollars). Table 12 presents the results for this outcome variable. The set of regressions include controls for time (year and week) and mall fixed-effects as well as for number of employees, number products sold at the store, weekend dummies, and region dummies.

The figures under Column (1) indicate that, on average, the stores where only store managers received training (T1) increased daily sales by 119 thousand Chilean pesos (176 dollars)²¹ compared with the stores in the control group. The intervention on both sales associates and managers (T2) also displays positive effects. When restricting the sample to the period in which only store managers had received the training (Column (2)), the estimated impact of the intervention is 110 thousand pesos. Column (3) presents the results for the

²⁰Thus, in the first case, we estimate the effect of a trained manager during the days between the end of the manager's training (June 22th) and the beginning of the team training (August 1st). In the second case, we are estimating the effect during the days covering the period after the team's training completion (approximately August 31st), and the end of the period of analysis (November 16, 2016).

²¹675 CLP = 1 US dollar.

post-treatment period when both managers and sales associates in *T2* stores had been treated. Here daily sales increase by 138 thousand pesos, equivalent to 204 dollars or 12.1% of average daily sales at the store-level. The comparison of columns (2) and (3) suggests the existence of spillovers between manager and sales associates within the stores. We explore this below. The intervention on only sales associates (*T3* or column 4) raise sales in 9 thousand pesos (13 dollars), but this effect is not statistically different from zero.²²

Overall, these findings indicate that the largest gains in productivity, proxied by daily store-level sales, emerge among stores where salespeople and store managers were trained. We obtain similar results if we estimate the model using the complete sample (all months in the pre-treatment period). Those results are presented in Table A1 in the Appendix.

The training intervention could have boosted sales by either fostering the teams' capacity to sell more expensive products (intensive margin), by increasing the number of total transactions (extensive margin), or by a combination of these two channels. To shed light on this matter, Table 13 displays the results obtained from the DD model using the number of daily transactions at the store level as the outcome variable. Our findings show overall positive impacts, but highlight the different mechanisms driving the increase in sales depending who receives training in the store.

As above, Column (1) presents the results when *T1* stores are compared to those in the control group. The large estimated effect implies that the training of only the store managers increased the number of transactions in approximately 6 events per day, a 11% increase over the average daily transactions observed during the period. Interestingly, the estimates for the stores where sales associates and manager received training (columns 2 and 3) are considerably smaller and only significant at 10%. Interestingly, although less clear than in Table 12, these findings also suggest the presence of spillovers between employees within the stores. When it comes to *T3* (Column D), we document a small positive effect of the intervention. However, the estimated value is not statistically different from zero.²³

²²Similar results were obtained using the information aggregated at a lower frequency -weeks, months, year-.

²³Using these results we estimate that the economic benefits of the intervention are \$1.38 million dollars (computed by multiplying the estimated effect on daily sales by the number of days since the end of the intervention and the number of treated stores). On the other hand, this intervention has also a detailed quantification of the costs of the training program, including human capital costs (technical teams in charge of the design, the training session, the costs of designing and implementing the skill surveys and other costs from the side of the firm, related to logistics and the opportunity costs of training. The overall cost was \$773,075. Using this information, we can conclude that after just three months, the rate of return is approximately 78%.

Table A2 presents results for the whole sample (expanding the sample to all weeks). These findings are similar in magnitude but less precisely estimated, suggesting the importance of excluding seasonal patterns characterizing the first weeks of the year. All in all, our results indicate positive effects of training on daily transactions, particularly for treatment groups $T1$ and $T2$.

7 Individual-level analysis

Even though ours is a store-based intervention, we can exploit the rich individual data to assess the impact of training on store managers and sales personnel. To do this, we estimate versions of equation (1) but using individual-level information instead of store-level data. We estimate:

$$Y_{i,t} = \beta_0 + \beta_1 D_i + \beta_2 B_t + \beta_3 D_i B_t + \beta_4 X_i + \beta_5 g_t + E_{i,t}, \quad (3)$$

where we use the sub-index i to denote individuals. The definition of the variables follows from our store-level model, thus $Y_{i,t}$ represents the outcome of interest of individual i in period t , D_i is a binary indicator taking a value of one if individual i received training and equals to zero if belonged to the control group, B_t defines the period (before or after training), X_i is a vector of individual-level controls measured at baseline (e.g., age and gender) including region fixed-effects, g_t is a set of time controls (month and year fixed-effects), and $E_{i,t}$ is the error term. The parameter of interest is again β_3 . The identification of this parameter is based on the comparison of individuals assigned to treatment stores who participated in the training and individuals assigned to control stores, before and after the training program.

Moreover, to provide a precise idea of the effect of the intervention, particularly the potential impact of communication skills among sales personnel and leadership among store managers, equation (3) is estimated using three different samples: overall (pooled sample of sales personnel and store managers), only sales personnel and only store managers. On the other hand, since the timing of the intervention varies across individuals, we adjust the definition of B_t accordingly. In particular, for the results obtained from the pooled regression we use the most restrictive definition (pre-treatment ends before the starting date for the managers training and post-treatment starts when sales personnel training ends). Finally, we follow our store-level analysis

and restrict the individual data to match the same months in post- and pre-treatment periods (between mid-June and the end of November).²⁴

Findings. We first present results using monthly sales as the dependent variable. These are displayed in Table 14. In the pooled regression (Column (1)), the estimated effect of training is 920,000 CLP (approximately 1,400 dollars). For sales associates (Column (2)) and managers (Column (3)), monthly sales increased by 810,000 CLP and 900,000 CLP, respectively. Thus, these results confirm our store-level findings. The intervention led to significant growth in sales across groups.

We next assess whether the previous findings are driven by changes at the extensive margin. To this end, Table 15 displays the results using individual’s total number of transactions per month as the dependent variable. Column (1) displays the results for the sample of all employees. We find that, on average, training increased the number of transactions per worker in 40 events per month. The result is statistically significant at 5%. Columns (2) and (3) report the effects for sales associates and managers, respectively. For the former group, the estimated effect is more than 50 extra transactions per month, whereas for the latter the impact is less than 20 transactions per month and non-statistically different from zero. These findings indicate the main impact came from the sales associates.

7.1 Spillovers

As described above, the experimental approach secured the random assignment of training to stores, but it did not limit the company’s capacity to re-allocate employees across stores.²⁵ As a result, individuals who were originally assigned to stores with an untrained manager (e.g., *T3* stores) could unexpectedly end up with a trained one. We exploit this phenomenon to investigate the presence of spillovers from trained managers/supervisors to untrained sales personnel. Formally, we estimate:

$$Y_{i,j,t} = \gamma_0 + \gamma_1 S_{it} + \gamma_2 B_t + \gamma_3 S_{it} B_t + \theta_i + \theta_j + E_{i,j,t} \quad (4)$$

²⁴The results using data for the whole period are presented in Tables A3 and A4 in the Appendix. By comparing these results, we can assess the robustness of our findings to pre-treatment trends and seasonality not captured by g_t .

²⁵The intervention was designed to modify the skill levels of those working at treated stores by October of 2015. However, after the original treatment assignment the firm was free to re-assign managers and sales personnel across stores.

where $Y_{i,j,t}$ is the outcome of interest of individual i in store j in period t , θ_i is an individual fixed-effect, θ_j is a baseline store fixed-effect and S_{it} is a binary indicator taking a value of one if individual i in period t was untrained and assigned to store j with a trained manager and equals to zero if belonged to a store in the control group, B_t defines the period (before or after the training of managers), $E_{i,j,t}$ is the error term and γ_3 is the parameter of interest.

The presence of individual and store fixed-effects represents a major difference between (4) and our previous regression models. In this case, θ_i and θ_j control for the potential strategic reallocation of workers across stores, isolating the impact of the intervention on untrained individuals. Therefore, one should use caution when comparing point estimates across equations as the source of identification depends on the sample and specification. In fact, while equation (4) exploits store switchers to identify the parameter of interest, our previous findings took advantage of the timing of the intervention and the treatment status of all individuals.²⁶

Table 16 presents our results. For sales (column (1)) and number of transactions (column (2)), we document positive and significant effects. Despite the fact the sources of identification differ across specifications, the magnitudes of these point estimates are similar to those reported in Tables 14 and 15. Consistent with our previous findings, positive spillovers from managers to sales personnel emerge as a driver of our results. Columns (3) and (4) repeat the analysis using only store. The results are similar in magnitude although less precise.

7.2 Exploring Mechanisms

Our findings suggest the training intervention increased treated units' productivity levels. However, they are silent about the specific mechanisms triggering the gains. For instance, rising total sales and/or transactions among "treated" store managers could have been the direct result of their improved leadership skills, but also the indirect consequence of higher levels of communication skills among the "treated" sales personnel working at the same stores. Likewise, increasing sales and transactions among sales personnel could have been the direct result of higher levels of communication skills within this group, but also the indirect consequence of higher levels of leadership among "treated" managers. Thus, two critical questions remain: Did the intervention modify skills? And if so, what skills?

²⁶Within the sample of untrained store associates, store- and individual-level fixed-effects can be added to the regression because treatment status can change over time. In fact, 63% of untrained sales associates worked at some point in a store with an untrained manager and approximately a 6% of untrained sales associates switched status within the period of reference.

To evaluate the relative importance of the alternative mechanisms behind the estimated effects on sales and transactions, we gather information from the two ability surveys (baseline and follow up) and estimate the regression model:

$$T_{i,t} = \delta_0 + \delta_1 D_i + \delta_2 B_t + \delta_3 D_i B_t + \xi_{i,t}, \quad (5)$$

where $T_{i,t}$ denotes the skill level of individual i as recorded in period t , and D_i and B_t are defined as before, $\xi_{i,t}$ is a vector of individual variables at the base line (i.e., age and gender).

Following the main objectives of the intervention, we use the scale of leadership as dependent variable for managers, and the measured of communication skills for store personnel. In both cases, the parameter of interest is δ_3 as it identifies the effect of the training on skills under the difference-in-difference strategy.

Table 17 presents the estimated impact of the intervention on skills based on the regression model (5). On average, as a result of the intervention, “treated” managers experience a 0.4 standard deviation increase in the leadership scale relative to the control group. Importantly, despite the small sample size (only 88 managers), the point estimate is significant at 10% level. On the other hand, the results for communication skills suggest a small and non-significant impact of the intervention among sales personnel. This result must be interpreted with caution as skill measures might not be precise proxies for true skills. Moreover, even though the intervention was aimed at boosting a specific skill, it could have modified other skill dimensions.

Next, we explore the extent to which the impact on sales reported among managers (column (3) in Table 14) can be explained by the 0.4 standard deviation increase in leadership attributed to the intervention. We do this by adding the interaction between treatment status, time period and leadership to the set of controls in equation 2. The results are reported in column (2) of Table 18, while column (1) repeat the original estimates. After controlling for leadership, the magnitude of the coefficients reduces 73%. Thus, a large fraction of the estimated impact can be linked to changes in leadership. Nevertheless, still more than half of the original point estimates remain unexplained. This, of course, could simply indicate the practical limitations of our skill measures. However, it could also suggest potential spillovers from sales personnel to store managers. Unraveling the relative importance of these alternatives should govern the efforts of the future research agenda on this subject.

8 Conclusions

This paper utilizes an experimental design combined with a difference-in-difference empirical strategy to quantify the impact of a training intervention intended to modify two socio-emotional skills - leadership and communication- among employees of one of the largest retailers in Latin America. Either at the store- or individual-level, the analysis delivers positive effects of the intervention on two proxies for productivity: sales and number of transactions. The following is a list of our main findings:

- From the store-level analysis we conclude the intervention increased daily sales by 119 dollars in locations where only managers were treated -*T1*- and by 204 dollars (12,1%) when both sales associates and managers received training -*T2*- (relative to control stores). The results are robust to different sample definitions and, in general, statistically significant at conventional levels (the exception is *T3*, i.e., the treatment group where only sales associates received training).
- Positive effects also emerge using total daily transactions as the outcome of interest, suggesting an impact on the extensive margin of sales. In this case, the largest estimated impact comes from stores in which only managers were assigned to training (11% increase in total transaction per day).
- To a large extent, our worker-level estimates confirm the previous findings. For sales, the estimated impact of the training program is unambiguously positive (and statistically significant) across the different treatment groups. For the number of monthly transactions, we document positive effects as well. Statistically significant results are obtained for sales associates and in the regression that pools both managers and sales associates. The training for both the managers and the sales associated had an unambiguous effect on the intensive margin. However, our results suggest that only the training for sales associates impact the extensive margin.
- Importantly, the individual analysis allows us to investigate the direct impact of the intervention on skills. While for sales associates we do not find any indication of gains in communication skills, among store managers we do document a positive and significant impact on leadership/management skills equivalent to 0.4 standard deviation.

- We estimate positive spillover effects from trained store managers to untrained sales personnel, which are consistent with our store-level results from *T1* stores (only managers received training). The managers training not only impacted their own sales but also had an indirect effect on the number of transactions and sales of the untrained workers in the stores. Since the identification comes from individuals switching stores after the training program was implemented, we mitigate potential threats to identification by controlling for individual- and store-level fixed effects.
- Our study highlights the important role of managers within the stores. Particularly, it shows that training them has both direct and indirect effects on productivity. Trained managers not only have large effects on untrained workers, but as shown by our store-level results, there might be complementarities with the enhanced skills of sales associates. These results are in line with the recent literature on the importance of managers to explain worker's productivity (Lazear et al., 2015) and on the effectiveness of management practices (Gosnell et al., 2019). Moreover, they highlight the multiple effects of manager's training on the productivity of the firm.
- Even though a large fraction of the impact on sales can be explained by the extra units of management skills, a significant proportion of the impact remains unexplained. We argue this can reflect both the limitations of our skill measures as well as the potential role of spillovers within stores. In this context, a future exploration of changes in the composition of the work force within each store during the post-treatment period might provide additional insights explaining the overall impact.

In sum, this paper shows not only how a tailored made training interventions can shape workers' performance but also the effectiveness of investing in soft-skills as they can trigger productivity gains. More generally, the analysis sheds light on three critical aspects to be considered for public policies aimed at improving labor market outcomes: the potential impact of public-private partnerships, the importance of institutional and managerial capacities, and the fundamental role of quality assurance mechanisms. Although the external validity of this type of study is not guaranteed, our findings suggest it is possible for a well-targeted on-the-job-training intervention to positively impact workers' productivity levels.

References

- Acemoglu, Daron and David Autor**, *Skills, tasks and technologies: Implications for employment and earnings*, Vol. 4, Elsevier B.V., 2011.
- Adhvaryu, Achyuta, Namrata Kala, and Anant Nyshadham**, “The Skills to Pay the Bills: Returns to On-the-job Soft Skills Training,” *NBER Working Paper No. W24313*, 2018.
- Almeida, Rita and Pedro Carneiro**, “The return to firm investments in human capital,” *Labour Economics*, 2009, 16 (1), 97–106.
- Almlund, Mathilde, Angela Lee Duckworth, James Heckman, and Tim Kautz**, “Personality psychology and economics,” in “Handbook of the Economics of Education,” Vol. 4, Elsevier, 2011, pp. 1–181.
- Barrett, Alan and Philip J. O’Connell**, “Does Training Generally Work? The Returns to In-Company Training,” *Industrial and Labor Relations Review*, April 2001, 54 (3), 647.
- Bassanini, Andrea, Alison L Booth, Giorgio Brunello, Maria De Paola, and Edwin Leuven**, “Workplace training in Europe,” *IZA Discussion Paper No. 1640*, 2005.
- Becker, Gary**, *Human Capital*, New York: Columbia University Press, 1964.
- Bloom, Howard S**, “Minimum detectable effects: A simple way to report the statistical power of experimental designs,” *Evaluation review*, 1995, 19 (5), 547–556.
- Borghans, Lex, Angela L. Duckworth, James J. Heckman, and Bas ter Weel**, “The Economics and Psychology of Personality Traits,” *Journal of Human Resources*, Fall 2008, 43 (4), 972–1059.
- Chetty, Raj, John N. Friedman, and Jonah E. Rockoff**, “Measuring the Impacts of Teachers II: Teacher Value-Added and Student Outcomes in Adulthood,” *American Economic Review*, September 2014, 104 (9), 2633–79.
- Conti, Gabriella**, “Training, productivity and wages in Italy,” *Labour Economics*, 2005, 12 (4), 557–576.

- Cunha, Flavio, James J. Heckman, Lance Lochner, and Dimitriy V. Masterov**, “Chapter 12 Interpreting the Evidence on Life Cycle Skill Formation,” *Handbook of the Economics of Education*, 2006, 1 (06), 697–812.
- Daruich, Diego**, “The Macroeconomic Consequences of Early Childhood Development Policies,” *Federal Reserve Bank of St. Louis Working Paper 2018-29*, 2018.
- De Grip, Andries and Jan Sauermann**, “The Effects of Training on Own and Co-Worker Productivity: Evidence from a Field Experiment,” *The Economic Journal*, 2011, 122 (560), 376–399.
- Dearden, Lorraine, Howard Reed, and John Van Reenen**, “The impact of training on productivity and wages: Evidence from British Panel Data,” *Oxford Bulletin of Economics and Statistics*, August 2006, 68 (4), 397–421.
- Deaton, Angus and Nancy Cartwright**, “Understanding and misunderstanding randomized controlled trials,” *Social Science and Medicine*, August 2018, 210, 2–21.
- Duflo, Esther, Rachel Glennerster, and Michael Kremer**, “Using Randomization in Development Economics Research: A Toolkit,” in T. Schultz and John Strauss, eds., *Handbook of Development Economic*, Vol. 4, Amsterdam and New York: North Holland, 2006, pp. 1051–1078.
- González-Velosa, Carolina, David Rosas, and Roberto Flores**, “On-the-Job Training in Latin America and the Caribbean: Recent Evidence,” in M. Grazzi and C. Pietrobelli (eds.), ed., *Firm Innovation and Productivity in Latin America and the Caribbean*, New York: Palgrave Macmillan US, 2016, chapter 5, pp. 137–166.
- Gosnell, Greer K, John A List, and Robert D Metcalfe**, “The Impact of Management Practices on Employee Productivity: A Field Experiment with Airline Captains,” Working Paper 25620, National Bureau of Economic Research February 2019.
- Hanushek, Eric A and Ludger Woessmann**, “Do better schools lead to more growth? Cognitive skills, economic outcomes, and causation,” *Journal of economic growth*, 2012, 17 (4), 267–321.

- Hanushek, Eric and Ludger Woessmann**, “The Role of Cognitive Skills in Economic Development,” *Journal of Economic Literature*, September 2008, 46 (3), 607–668.
- Heckman, James J.**, “Detecting Discrimination,” *Journal of Economic Perspectives*, Spring 1998, 12 (2), 101–116.
- and **Tim Kautz**, “Hard evidence on soft skills,” *Labour Economics*, 2012, 19 (4), 451–464.
- , **J. Stixrud**, and **S. Urzúa**, “The effects of cognitive and noncognitive abilities on labor market outcomes and social behavior,” *Journal of Labor Economics*, 2006, 24 (3), 411–482.
- , **Robert J. Lalonde**, and **Jeffrey A. Smith**, “Chapter 31 -The Economics and Econometrics of Active Labor Market Programs,” in Orley C. Ashenfelter and David Card, eds., *Orley C. Ashenfelter and David Card, eds., Vol. 3 of Handbook of Labor Economics*, Elsevier, 1999, pp. 1865 – 2097.
- Ibarrarán, Pablo, Alessandro Maffioli, and Rodolfo Stucchi**, “SME Policy and Firms’ Productivity in Latin America,” *IZA Discussion Paper*, 2009, (4486).
- Ibarrarán, Pablo, Jochen Kluve, Laura Ripani, and David Rosas**, “Experimental Evidence on the Long-Term Impacts of a Youth Training Program,” *IZA Working Paper*, 2015, (9136).
- Ibarraran, Pablo, Laura Ripani, Bibiana Taboada, Juan M. Villa, and Brigida García**, “Life skills, employability and training for disadvantaged youth: Evidence from a randomized evaluation design,” *IZA Journal of Labor Development*, 2014, 3(1), 1–24.
- Kern, Margaret, Angela Duckworth, Sergio Urzúa, Rolf Loeber, Magda Stouthamer-Loeber, and Donald Lynam**, “Do as you’re told! Facets of agreeableness and early adult outcomes for inner-city boys,” *Journal of Research in Personality*, 2013, 47 (6), 795–799.
- Kluve, J.**, “A review of the effectiveness of Active Labour Market Programmes with a focus on Latin America and the Caribbean. Research department working paper no. 9,” 2016.
- Konings, Jozef and Stijn Vanormelingen**, “The Impact of Training on Productivity and Wages: Firm-Level Evidence,” *Review of Economics and Statistics*, May 2015, 97(2), 485–497.

Lazear, Edward P, Shaw Kathryn L., and Christopher T Stanton, “The Value of Bosses,” *Journal of Labor Economics*, 2015, 33 (4), 823–861.

Mitnik, Oscar A, Laura Ripani, and David Rosas-Shady, “Comparing the Results of Youth Training Programs in Latin America and the Caribbean,” Technical Report, Inter-American Development Bank 2016.

Prada, María F. and Sergio Urzúa, “One Size Does Not Fit All: Multiple Dimensions of Ability, College Attendance, and Earnings,” *Journal of Labor Economics*, 2017, 35 (4), 953–991.

Prada, María F. and Graciana Rucci, “Instrumentos para la medición de las habilidades de la fuerza de trabajo”, Technical Note # 1070, July 2016, Inter-American Development Bank, Washington, DC.

Roberts, Brent W, Nathan R. Kuncel, Rebecca Shiner, Avshalom Caspi, and Lewis R. Goldberg, “The power of personality: The comparative validity of personality traits, socioeconomic status, and cognitive ability for predicting important life outcomes,” *Perspectives on Psychological science*, 2007, 2 (4), 313–345.

Urzúa, Sergio, “Racial Labor Market Gaps,” *Journal of Human Resources*, 2008, 43 (4), 919–971.

Urzúa, Sergio and Esteban Puentes, “La evidencia del impacto de los programas de capacitación en el desempeño en el mercado laboral,” Technical Note # 268, Dec 2010, Inter-American Development Bank.

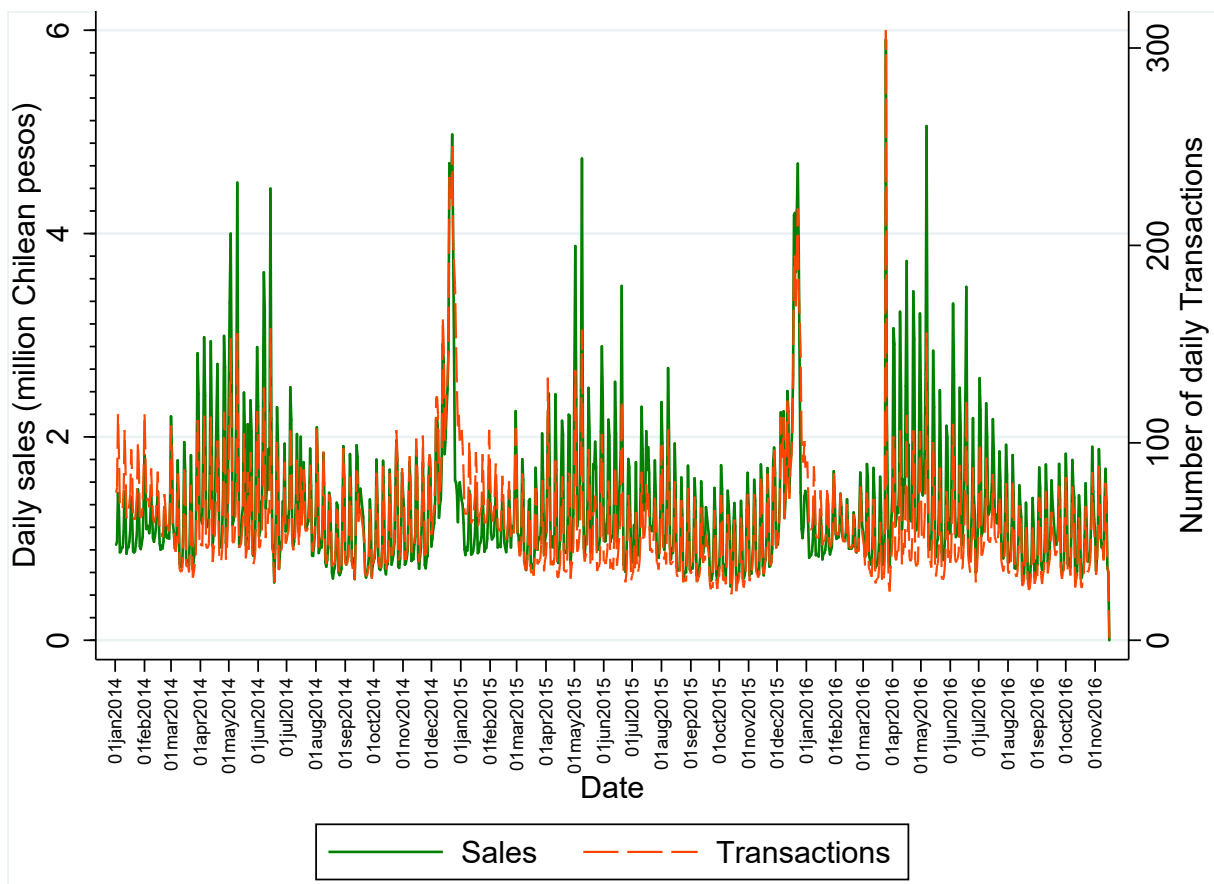


Figure 1: Average Daily Store Sales and Daily Store Transactions, Period 2014-2016

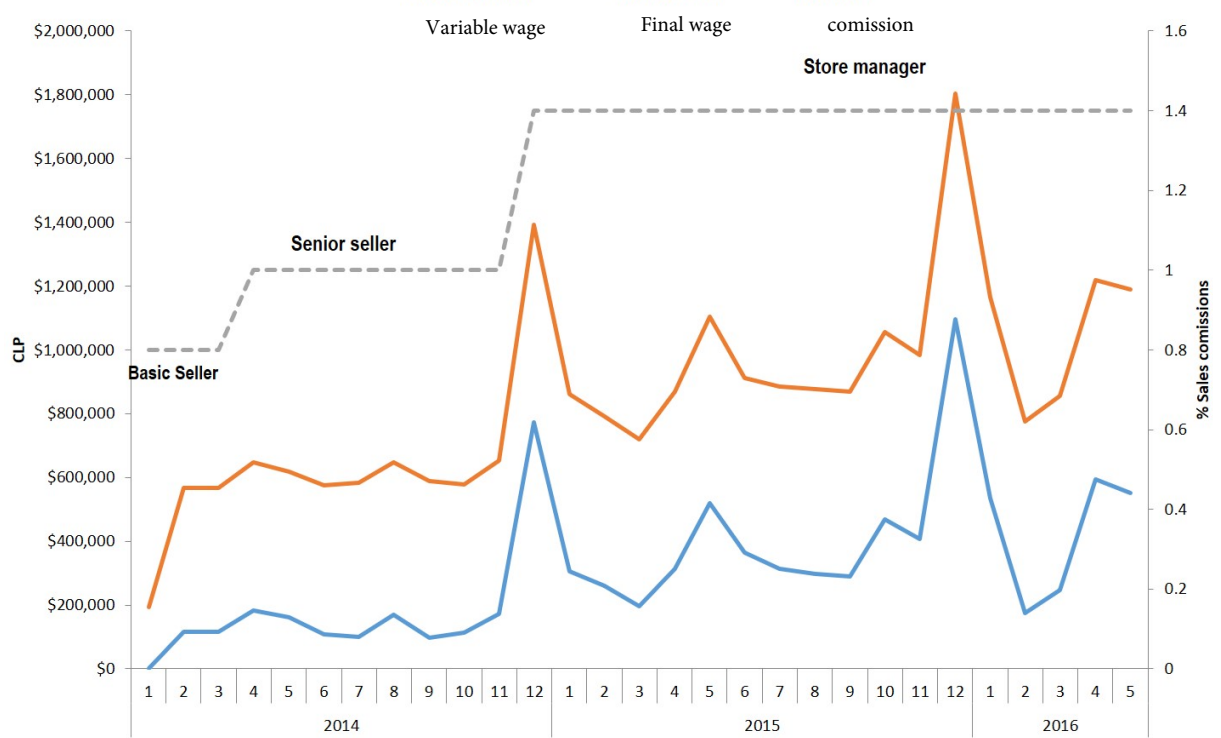


Figure 2: Within firm progression: Example from an employee, 2014-2016

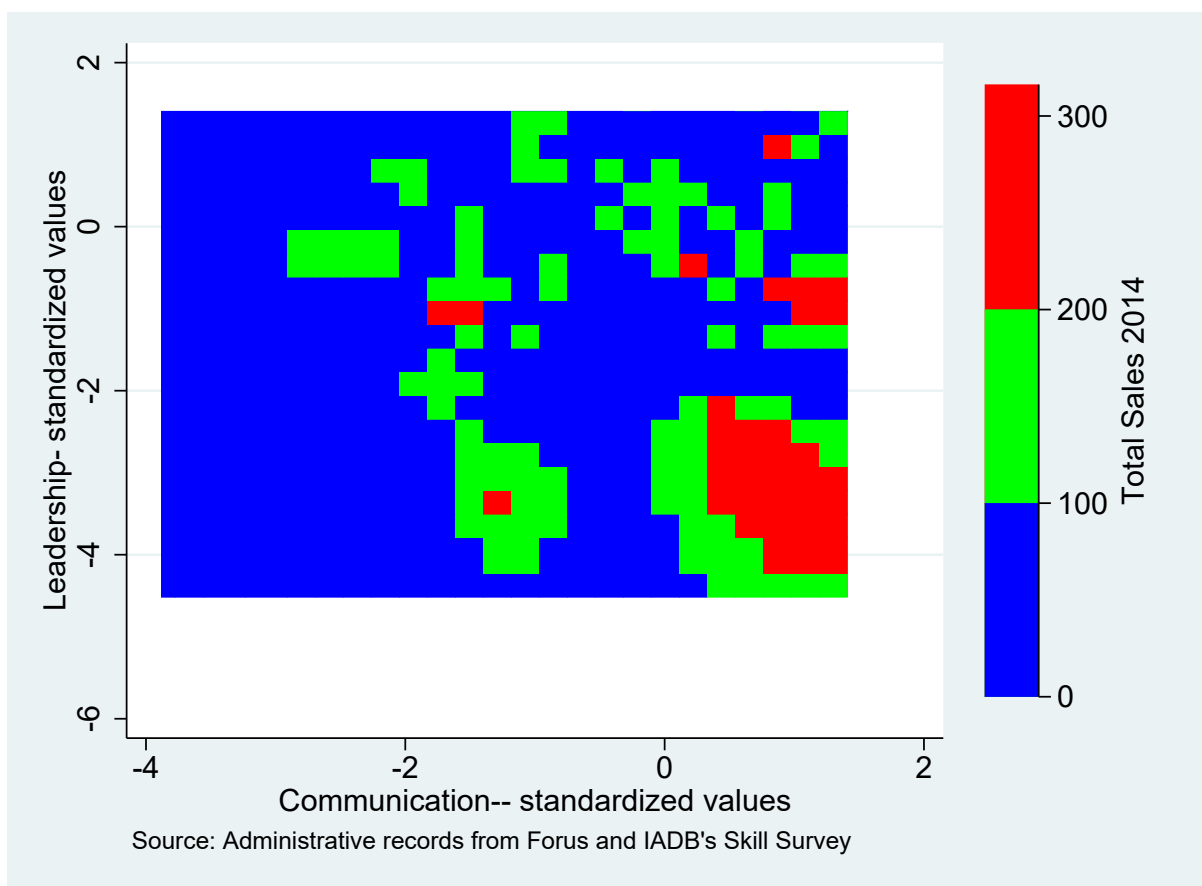


Figure 3: Total Sales, Communication and Leadership Skills for Sales Associates - 2014

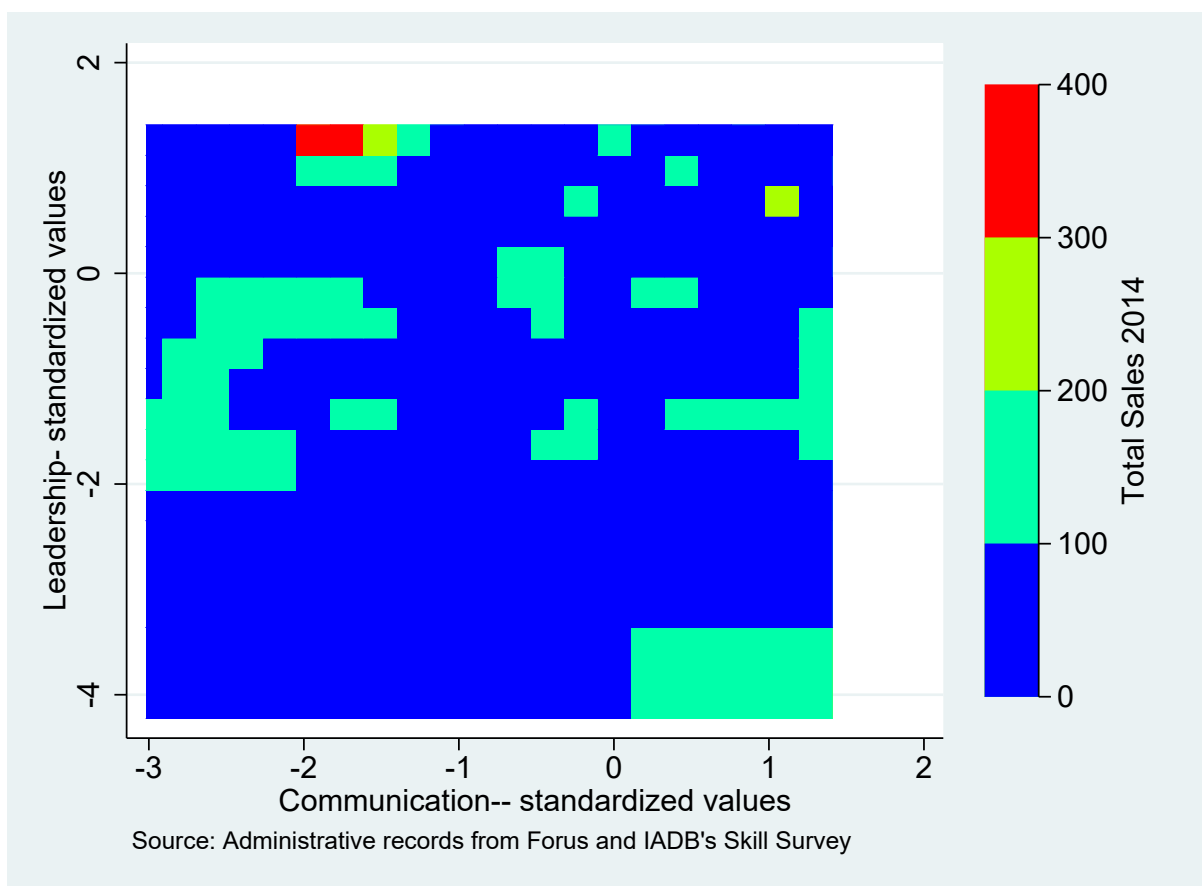


Figure 4: Total Sales, Communication and Leadership Skills for Managers - 2014

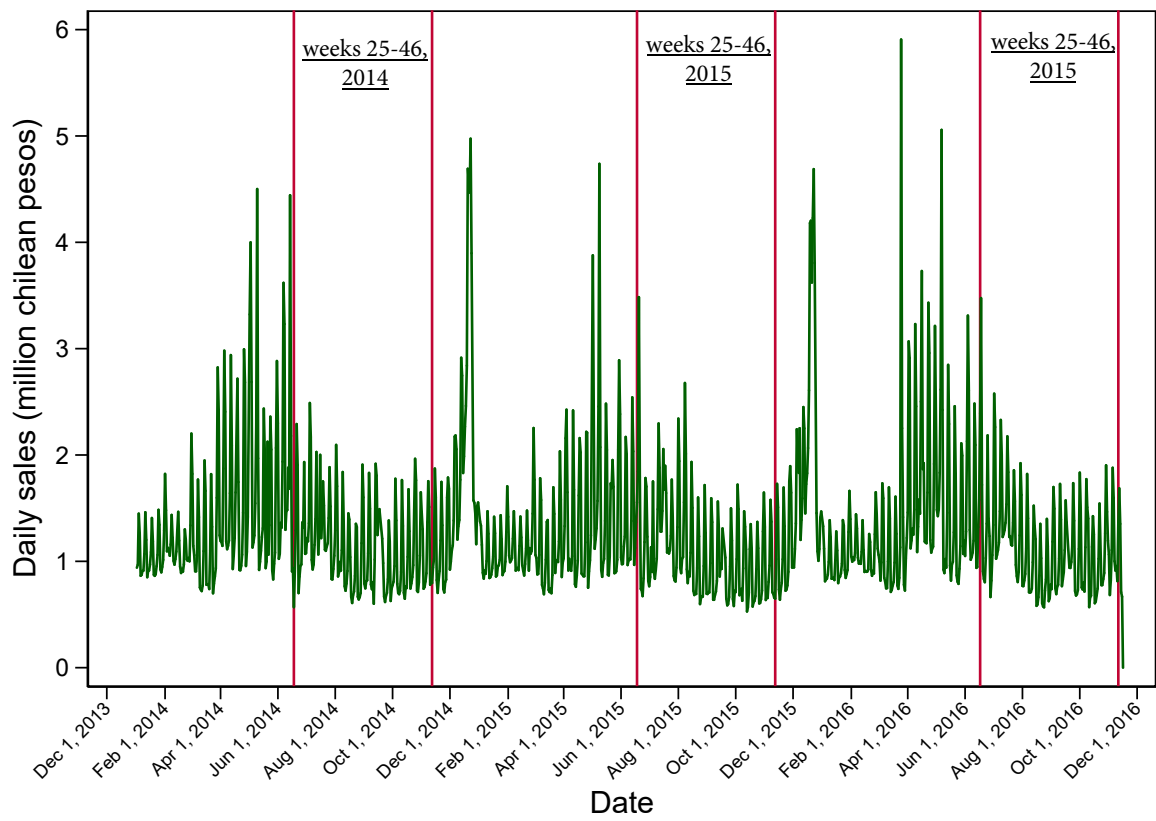


Figure 5: Average Daily Store Sales, weeks 25-46, 2014-2016

Table 1: Modules and objectives- Management program

Day	Objectives
Day 1 Managing yourself	Enhance managers self-knowledge and understand how they control their own personal effectiveness. This interactive day provides the managers with the tools to help manage problems in the store, encourage teamwork and become a role model for their team. Leadership and management skills.
Day 2 Managing communication and influencing	Enhance the managers personal effectiveness with communication and influencing techniques, this day focuses on effective communication skills (the use of positive language, emphatic listening) that can be used to build effective relationships, achieve targets and motivate staff.
Day 3 Managing others	Enhance the managers personal effectiveness in communicating and managing others through effective delegation, conflict management and negotiation.
Day 4 Managing and achieving companys targets	Enhance and understand the role of assertiveness and clear focus required by the Manager to achieve companys targets, this session helps the Managers develop techniques to achieve business goals whilst dealing with people and situations. Confidence, self-control and decision making in difficult times, communication skills to empower the members of the team, tools to learn how to prioritize individual and team work to achieve business goals.
Day 5: Coaching for Success	Learn the theory behind coaching and how to use it to influence team members and identify individual potential Develop the managers coaching skills, providing the tools and techniques required to identify different types of staff and how to motivate and train them.

Table 2: Modules and objectives: Team program

Module	Objectives
1: To serve I	Understanding what customer service excellence means at the company, the benefits of delivering it and the behaviors and skills needed to meet companys expectations.
2: To serve II	Knowing your customers, understanding companys store and brands, creating impact through strong first impressions of you and your store (non-verbal communication)
3: To understand I	Greeting your customer, identifying the needs and meeting the expectations of your customer, being an ambassador for your store. Applied communication skills.
4: To understand II	Use communication skills to show empathy and looking after customer concerns, and to deal with conflict situations
5. To carry out I	Understanding your role in sales and what is selling, influencing the customer and the sales cycle
6. To carry out II	The selling process at the company
7. To sell	Understanding and communicating technical details of the products sold to customers.
8. To impact	Recognizing internal and external clients and adapting communication styles to different costumers
9. To communicate I	Contact points in the customers journey, positive talking to customers
10. To communicate II	Communication techniques and tools to be confident. Communication with different customers and cultures.
11. To identify	Learn about the selling process (4ps) and the role of communication
12. To optimize your service	Develop a personal Action plan and your assessment

Table 3: Stores by geographic location and mall restrictions: Overall versus program's participants

	(1) Total stores	(2) In shopping mall	(3) Large mall*	(4) % in mall (2)/(1)
(A) Country	324	246	216	76%
(B) Regions V, VIII and R.M.	222	176	155	79%
(C) 5 selected cities	193	157	138	81%
(C)/(A) $\times 100$ %	60%	63%	64%	

Note: The program was implemented in the three largest regions of Chile (V, VIII and RM). See Section 3 for further details on the experimental design. *Large mall refers to a shopping mall that has three or more of the firm stores.

Table 4: Intended and Final Treatment Groups: Number of stores

Intended treatments	Final treatment groups						Total Eligible
	T1	T2	T3	C	Total	Non Compliers	Total
T1	33	0	0	9	42	0	42
T2	1	26	13	0	40	2 ^a	42
C	3	0	0	50	53	1 ^a	54
Total	37	26	13	59	135	3	138

Note: (a) Closed stores.

Table 5: Programmed training vs. Actual training: Managers and Sales associates

	Planned to receive training (1)			Actually trained (2)		
	Managers	Sales associates	Total	Managers	Sales associates	Total
T1	42		0	33	1	34
T2	40		179	25	62	87
C	0		0	3	6	9
New Hires				10	107	117
Total	82		179	71	176	247
% trained (2)/(1) $\times 100$ %				87%	98%	95%

Note: The first panel of the table "Planned to receive training (1)" reflects the original plan according to the randomization process and the number of employees at that time. The information of the second panel of the table "Actually trained (2)" comes from the firm's records of the number of participants that attended and satisfactorily complete training.

Table 6: Balance tests: Differences between groups in Baseline Variables $t - 1$

Variables	T1 vs C		T2 vs C		T3 vs C		T1 vs T2		T1 vs T3		T2 vs T3	
	Dif.	t-stat	Dif.	t-stat	Dif.	t-stat	Dif.	t-stat	Dif.	t-stat	Dif.	t-stat
Region V	0.02	0.66	-0.06	1.00	0.14	1.03	-0.06	1.17	-0.12	1.06	-0.20	1.10
Region VIII	-0.04	1.11	-0.13	1.37	0.20	1.18	-0.12	1.37	-0.24	1.42	-0.32	1.32
Region XIII	0.02	0.37	0.19	1.76	-0.33	2.44	0.19	1.85	0.35	2.75	0.52	2.43
Weekend dummy	0.00	0.04	0.00	0.49	0.00	1.54	0.00	0.17	0.00	1.49	0.00	0.65
Variety	-0.01	0.16	-0.06	0.91	0.05	0.45	-0.09	1.30	-0.07	0.47	-0.14	1.23
Brand	0.20	0.27	-0.80	0.64	0.17	0.23	-0.72	0.58	0.03	0.03	-0.89	0.97
Brand (short)	0.13	1.22	0.03	0.34	-0.04	0.20	-0.07	0.60	0.18	0.80	-0.02	0.09
Store sales (CLP)	59,905	0.38	-80,323	0.46	46,617	0.21	-239,330	1.21	13,845	0.06	-285,553	1.05
Sales per worker (CLP)	11,419	0.92	-7,908	0.36	17,996.3	0.71	-13,144	0.60	-6,484	0.23	-31,065	0.91
Sales per transaction (CLP)	1,572	1.15	-2,793	1.53	2,325	0.69	-2,437	1.34	(749.7)	0.26	-4,765.2	1.67
Number of transactions	-3.77	0.51	1.57	0.22	-5.73	0.67	-8.78	0.92	1.99	0.23	-3.01	0.37
Transactions per worker	-0.23	0.38	0.44	0.50	-0.11	0.12	-0.01	0.01	-0.11	0.11	0.11	0.11
N obs.	79,687		67,405		70,488		59,811		41,302		32,103	

Note: Test of differences calculated using errors clustered at the mall level according to the experimental design. The statistical mean difference tests are generated using information for the years 2014 and 2015. CLP: Chilean Pesos. As a reference, 675 CLP=1 US dollar. Region V, VIII and XIII are dummy variables taking the value of one if the store is located in Valparaíso, Concepción y Metropolitan Region (RM), respectively and zero otherwise. The variable "Variety" is an indicator of the degree of specialization of the store, it records how many product categories are sold at the store, where the categories are: footwear, clothing, accessories and others. The variables "Brand" and "Brand-Short" identify the main brands sold at each store. The "Brand" has 15 categories, while the short version of the variable collapses the brands into four categories depending on the average sale price per item.

Table 7: Summary Statistics: 2014-2016

Full sample	Controls		T1: Only managers		T2: Managers & sales associates		T3: Only sales associates	
	Mean	SD	Mean	SD	Mean	SD	Mean	SD
Outcome variables	1,328,237	1,433,751	1,399,018	1,254,789	1,104,988	1,036,414	1,371,626	1,089,416
	176,642	151,582	190,787	156,229	166,835	143,360	195,158	144,648
	22,950	12,629	24,299	10,959	20,310	9,672	25,015	10,730
	63.25	68.72	60.80	51.74	55.76	43.58	58.45	44.35
	8.34	6.49	8.33	6.23	8.55	6.21	8.38	5.92
Independent variables								
	7.10	2.72	7.20	1.87	6.53	1.58	6.94	1.37
	0.29	0.45	0.29	0.45	0.29	0.45	0.29	0.45
	0.05	0.23	0.08	0.28			0.20	0.40
	0.09	0.29	0.08	0.28			0.32	0.47
	0.85	0.35	0.83	0.37	1.00		0.48	0.50
	25.67	14.51	25.82	14.48	25.75	14.55	25.67	14.50
	2.64	0.51	2.63	0.50	2.56	0.53	2.70	0.47
	8.14	4.58	8.47	4.30	7.57	3.97	8.41	3.17
	1.49	0.60	1.66	0.47	1.46	0.50	1.48	0.50
N. obs.	56,373		37,006		25,905		12,970	
N. stores	59		37		26		13	

Note: CLP: Chilean Pesos. As a reference, 675 CLP=1 US dollar. Region V, VIII and XIII are dummy variables taking the value of one if the store is located in Valparaíso, Concepción y Metropolitan Region (RM), respectively and zero otherwise. The variable "Variety" is an indicator of the degree of specialization of the store, it records how many product categories are sold at the store, where the categories are: footwear, clothing, accessories and others. The variables "Brand" and "Brand-Short" identify the main brands sold at each store. The "Brand" has 15 categories, while the short version of the variable collapses the brands into four categories depending on the average sale price per item.

Table 8: Summary Statistics Individual-level: Skills and Main Outcomes

Variables	Mean	Std Dev	Min	Max	N obs	N ind
Skills in baseline survey t (t-1)						
Communication	0.02	0.98	-4.10	1.52	40,159	1,661
Leadership	0.03	0.99	-4.52	1.55	40,159	1,661
Metacognitive	0.00	0.98	-5.96	1.55	38,940	1,611
Self-efficacy	0.01	0.99	-3.63	4.18	37,506	1,549
Grit	0.02	0.99	-6.42	1.04	40,159	1,661
Openness	0.01	1.00	-6.50	0.96	40,159	1,661
Conscientiousness	0.04	0.99	-5.14	1.56	40,159	1,661
Extraversion	-0.02	1.01	-4.18	1.47	40,159	1,661
Agreeableness	-0.01	1.00	-4.70	1.46	40,159	1,661
Neuroticism	0.02	0.99	-2.52	3.09	40,159	1,661
Numeracy	0.04	0.99	-2.17	0.95	40,159	1,661
Functional Literacy	0.00	1.01	-3.75	0.27	40,159	1,661
Skills in follow up survey (t+1)						
Communication	-0.01	1.03	-4.34	1.52	30,932	1,569
Leadership	0.05	1.03	-4.75	1.64	30,932	1,569
Metacognitive	0.02	0.99	(6.91)	1.52	29,819	1,516
Self-efficacy	0.00	1.01	(7.48)	4.13	28,761	1,452
Grit	0.01	1.03	(6.34)	1.00	30,932	1,569
Openness	0.03	0.96	(6.22)	0.90	30,932	1,569
Conscientiousness	0.06	1.00	(5.43)	1.47	30,932	1,569
Extraversion	(0.07)	1.00	(4.20)	1.50	30,932	1,569
Agreeableness	(0.01)	1.00	(4.81)	1.44	30,932	1,569
Neuroticism	0.03	0.99	(2.37)	2.96	30,932	1,569
Numeracy	0.07	0.96	(2.16)	0.91	30,932	1,569
Functional Literacy	0.03	0.95	(3.46)	0.29	30,932	1,569
Other Variables of Interest 2014-2016						
Monthly Sales	7.09	6.03	0.00	87.90	32,383	1,373
No Transactions	272.99	251.82	1.00	5085.00	32,383	1,373
Training	0.22	0.41	0.00	1.00	23,795	1,141
Treatment_store in t-1	1.00	1.20	0.00	3.00	40,159	1,661
Treatment_store in t	1.83	1.06	0.00	3.00	31,892	2,589

Note: The baseline skills survey was collected during the months of November and December 2015. The follow up survey between November and December of 2016, with few responses in January of 2017. The other variables of interest cover the whole period between January 2014 and November 2016.

Table 9: Baseline leadership and communication skill levels as determinants of promotions during 2015

Variable	(1)	(2)	(3)
Leadership	0.200**	0.151*	0.162*
Communication	0.007	-0.065	-0.078
N. of Obs.	790	790	790
Controls			
Gender and experience	Y	Y	Y
Other socio-emotional skills	N	Y	Y
Cognitive skills	N	N	Y

Source: Authors own calculations based on firms information. Leadership and communication skills are standardized measures (mean=0 and variance=1). Sample includes only salespersons (as opposed to store managers) before the promotion. Inference based on clustered standard errors (store-level). ***: 1%, **: 5%, *: 10%.

Table 10: Baseline Leadership and Communication Skill as Determinants of Total Annual Sales in 2015.

Variables	Pooled		Sales Associates		Managers	
	(1)	(2)	(3)	(4)	(5)	(6)
Leadership	-0.18 (2.13)	-1.56 (2.48)	-0.76 (2.66)	-2.73 (3.14)	18.93* (9.72)	15.50 (13.29)
Communication	5.74*** (2.12)	4.62*** (2.33)	5.92*** (2.63)	5.45* (2.91)	-14.52 (8.63)	-22.91* (11.96)
Observations	1,173	1,173	916	916	257	257
R-squared	0.41	0.42	0.47	0.47	0.83	0.86
Controls						
Gender & Experience	Y	Y	Y	Y	Y	Y
Position	Y	Y	Y	Y	Y	Y
Socio-emotional skills	N	Y	N	Y	N	Y
Cognitive Skills	N	Y	N	Y	N	Y

Source: Authors own calculations based on firms information. Leadership and communication skills are standardized measures (mean=0 and variance=1). Inference based on clustered standard errors (store-level). ***, 1%, **, 5%, *, 10%

Table 11: Summary Statistics: Weeks 25-46

Weeks 25-46		Controls		T1: Only managers		T2: Managers & sales associates		T3: Only sales associates	
Variable		Mean	SD	Mean	SD	Mean	SD	Mean	SD
Outcome Variables									
Total daily store sales (CLP)		1,137,428	1,064,322	1,208,916	1,029,819	954,439	824,250	1,172,217	852,781
Daily sales per worker (CLP)		157,618	123,549	166,021	128,083	145,704	115,778	169,532	116,430
Daily sales per transaction (CLP)		22,936	10,918	24,109	9,827	20,283	9,021	24,707	9,819
Total daily transactions		51.79	47.25	51.73	41.90	47.76	34.83	49.86	35.35
Daily transactions per worker		7.21	5.14	7.17	5.14	7.41	5.07	7.31	5.03
Number of workers		6.96	2.14	7.14	1.71	6.48	1.53	6.86	1.27
Independent Variables									
Weekend dummy		0.28	0.45	0.28	0.45	0.28	0.45	0.28	0.45
Region V		0.07	0.25	0.08	0.28	-	-	0.21	0.41
Region VIII		0.09	0.29	0.08	0.27	-	-	0.32	0.47
Region XIII		0.84	0.37	0.84	0.37	1.00	-	0.47	0.50
Week		35.41	6.35	35.40	6.35	35.46	6.36	35.41	6.35
Variety		2.62	0.52	2.62	0.51	2.55	0.53	2.69	0.47
Brand		8.14	4.27	8.48	4.29	7.59	3.95	8.37	3.19
Brand_short		1.53	0.54	1.65	0.48	1.47	0.50	1.47	0.50
N. observations		24,844		16,439		11,448		5,731	

Notes: CLP: Chilean Pesos. As a reference, 675 CLP=1 US dollar. Region V, VIII and XIII are Vina, Concepcion y Metropolitan region, respectively. The variable "Variety" is an indicator of the degree of specialization of the store, it records how many product categories are sold at the store, where the categories are: footwear, clothing, accessories and others. The variables "Brand" and "Brand-Short" identify the main brands sold at each store. The "Brand" has 15 categories, while the short version of the variable collapses the brands into four categories depending on the average sale price per item.

Table 12: The impact on daily store sales (CLP)
Pre-treatment Sample: Only weeks 25-46 of years 2014 and 2015

Controls	T1: Only managers (1)	T2: Managers and Sales Associates (2)	T3: Sales Associates (4)
D_i	-93,192.9 (79,640.1)	-7,122.5 (61,908.2)	1,871.4 (97,857.8)
B_t	-27,769.3 (50,786.0)	-24,014.5 (69,648.1)	-167,644.8** (140,992.25)
Parameter of interest ($D_i \times B_t$)	118,812.8** (49,166.3)	110,905.1* (56,263.2)	137,979.5** (64,799.2)
Year/Week FE	Y	Y	Y
Mall FE	Y	Y	Y
Observations	40,757	35,829	28,467
R-squared	0.51	0.54	0.55
Treated Group	T1	T2	T3
Post-treatment Period (B_t^k)	–	Only managers have been trained	All store employees have been trained

Source: Authors own calculations based on firms information. As a reference, 675 CLP=1 US dollar. Regressions include the following set of controls: Number of store's employees, number of products sold at the store, weekend dummies, and region dummies. Inference based on clustered standard errors (mall-level). ***: 1%, **: 5%, *: 10%

Table 13: The impact on the number of daily transactions
Pre-treatment Sample: Only weeks 25-46 of years 2014 and 2015

	T1: Only managers	T2: Managers and Sales Associates	T3: Sales Associates
Controls	(1)	(2)	(3)
D_i	-3.52 (3.11)	-2.76 (2.12)	-2.28 (2.24)
B_t	-6.74*** (2.11)	-8.65** (3.58)	2.51 (3.67)
Parameter of interest ($D_i \times B_t$)	6.29** (2.86)	4.29* (2.19)	4.82* (2.65)
Year/Week FE	Y	Y	Y
Mall FE	Y	Y	Y
Observations	40,757	35,829	33,781
R-squared	0.58	0.60	0.59
Treated Group	T1	T2	T3
Post-treatment Period (B_t^k)	–	Only managers have been trained	All store employees have been trained

Source: Authors own calculations based on firms information. As a reference, 675 CLP=1 US dollar. Regressions include the following set of controls: Number of store's employees, number of products sold at the store, weekend dummies, and region dummies. Inference based on clustered standard errors (mall-level). ***, 1%; **, 5%; *, 10%

Table 14: The effect of the intervention on monthly sales (Millions of CLP)
Individual Level Analysis

Controls	Pooled (1)	Sales associates (2)	Managers (3)
D_i	-0.65* (0.36)	0.39 (0.48)	-0.30 (0.46)
B_t	-0.88*** (0.27)	-1.37*** (0.46)	-0.88 (0.67)
Parameter of interest ($D_i \times B_t$)	0.92** (0.40)	0.81* (0.45)	0.90* (0.48)
Observations	1,828	1,073	1,345
R-squared	0.05	0.11	0.10

Source: Authors own calculations based on firms information. Regressions include age at baseline and gender as controls as well as month, year and region fixed-effects. Inference based on clustered standard errors (mall-level). B Is defined so that the pre-treatment period includes information from 2014 and 2015 but only in the same months observed after treatment (September and October for sales associates and July to October for Managers). As a reference, 675 CLP=1 US dollar. ***: 1%, **: 5%, *: 10%

Table 15: The effect of the intervention on total number of transactions
Individual Level Analysis

Controls	Pooled (1)	Sales associates (2)	Managers (3)
D_i	-30.30* (16.74)	5.29 (17.64)	-4.35 (19.36)
B_t	-60.75*** (14.11)	-84.23*** (20.84)	-60.37** (25.24)
Parameter of interest ($D_i \times B_t$)	40.57** (16.82)	52.91** (24.40)	19.06 (18.19)
Observations	1,828	1,167	1,345
R-squared	0.04	0.10	0.12

Source: Authors own calculations based on firms information. Regressions include age at baseline and gender as controls as well as month, year and region fixed-effects. Inference based on clustered standard errors (mall-level). B Is defined so that the pre-treatment period includes information from 2014 and 2015 but only in the same months observed after treatment (September and October for sales associates and July to October for Managers) ***: 1%, **: 5%, *: 10%

Table 16: Spillovers from Trained Store Managers to Untrained Sales Associates
Individual Level Analysis

Controls	All Sales Associates without Training			
	Sales	Transactions	Sales	Transactions
	CLP	#	CLP	#
	(1)	(2)	(3)	(4)
S_{it}	0.52 (0.61)	40.28* (23.58)	0.65 (1.00)	38.42 (48.01)
B_t	-1.39*** (0.25)	-71.13*** (9.78)	-1.17*** (0.35)	-66.85*** (17.97)
Parameter of interest ($S_{it} \times B_t$)	1.28*** (0.41)	51.50*** (15.97)	1.21* (0.73)	48.95* (24.94)
Individual FE	Y	Y	N	N
Store at baseline FE	Y	Y	Y	Y
Observations	2,040	2,040	1,585	1,585
R-squared	0.10	0.06	0.42	0.39
Number individuals	332	332		

Source: Authors own calculations based on firms information. In columns (3) and (4) we control for age at baseline and gender and cluster standard errors at mall level. ***: 1%, **: 5%, *: 10%

Table 17: The effect of the intervention on skills

Controls	Leadership	Communication
	Managers	Sales Associates
	(1)	(2)
D_i	-0.32 (0.25)	-0.02 (0.20)
B_t	-0.24 (0.19)	-0.00 (0.08)
Parameter of interest ($D_i \times B_t$)	0.40* (0.23)	-0.04 (0.17)
Observations	176	439
R-squared	0.16	0.13

Source: Authors own calculations based on firms information. Variables are standardized to have mean zero and SD equal to one. Inference based on clustered standard errors (mall-level) ***: 1%, **: 5%, *: 10%

Table 18: The effect of the intervention on monthly sales
Individual Level Analysis - Sample of Store Managers

Controls	Original (1)	Including Leadership (2)
D_i	-0.30 (0.46)	-0.15 (0.42)
B_t	-0.88 (0.67)	-0.57 (0.70)
Parameter of interest ($D_i \times B_t$)	0.90* (0.48)	0.73 (0.56)
$D_i \times B_t \times \text{Leadership}$	- -	0.22 (0.44)
Mall FE	Y	Y
B: Same months 2014 and 2015	Y	Y
Observations	1,345	1,254
R-squared	0.10	0.11

Source: Authors own calculations based on firms information. Regressions include age at baseline and gender as controls as well as month, year and region fixed-effects. Inference based on clustered standard errors (store-level) when more than one individual per store is included in the regressions. *B: Same months 2014 and 2015* indicates that pre-treatment period includes information from 2014 and 2015 but only in the same months observed after treatment (July to October for Managers). ***: 1%, **: 5%, *: 10%

Table A1: The impact on daily store sales (CLP)
Store Level Analysis

Controls	T=1 vs Control (A)	T=2 vs Control (B)	T=2 vs Control (C)	T=3 vs Control (D)
D_i	86,529.8 (112,397.2)	124,850.7 (137,579.5)	135,010.3 (140,071.2)	45,795.8 (227,075.6)
B_t	-28,823.0 (71,745.8)	-3,223.7 (64,380.4)	238,949.5*** (64,173.04)	-313,917.8*** (107,873.5)
Parameter of interest ($D_i \times B_t$)	70,635.8 (73,873.6)	106,100.88* (59,420.8)	132,323.3* (63,684.3)	16,702.5 (70,367.5)
Year/Week FE	Y	Y	Y	Y
Mall FE	Y	Y	Y	
Observations	90,553	76,787	76,765	67,235
R-squared	0.48	0.41	0.41	0.09
Post treatment definition	Only manager completed training	Only manager completed training	Manager and Sales associates completed training	Only sales associates completed training

Source: Authors own calculations based on firms information. As a reference, 680 CLP=1 US dollar (September 2016). Regressions include the following set of controls: Number of store's employees, number of products sold at the store, weekend dummies, and region dummies. Inference based on clustered standard errors (mall-level). ***, **, *, 1%, 5%, 10%

Table A2: The impact on the number of daily transactions
Store Level Analysis

Controls	T=1 vs Control (1)	T=2 vs Control (2)	T=2 vs Control (3)	T=3 vs Control (4)
D_i	-6.81 (5.09)	-1.21 (6.23)	-0.33 (6.03)	-0.37 (9.76)
B_t	4.52 (5.36)	4.38 (4.17)	2.35 (4.21)	3.42 (3.76)
Parameter of interest ($D_i \times B_t$)	5.88 (3.73)	6.28* (3.20)	5.59 (3.35)	0.88 (2.88)
Year/Week FE	Y	Y	Y	Y
Mall FE	Y	Y	Y	Y
Observations	90,553	79,790	79,767	67,235
R-squared	0.54	0.52	0.52	0.57
Post treatment definition	Only manager completed training	Only manager completed training	Manager and Sales associates completed training	Only sales associates completed training

Source: Authors own calculations based on firms information. Regressions include the following set of controls: Number of store's employees, number of products sold at the store, weekend dummies, and region dummies. Inference based on clustered standard errors (mall-level). ***, 1%, **, *, 10%

Table A3: The effect of the intervention on monthly sales (Millions of CLP)
Individual Level Analysis

Controls	Pooled (1)	Sales associates (2)	Managers (3)
D_i	-0.69 (0.43)	0.19 (0.55)	0.02 (0.61)
B_t	-0.84*** (0.30)	-0.83*** (0.28)	-1.32** (0.54)
Parameter of interest ($D_i \times B_t$)	1.01** (0.43)	1.11** (0.44)	0.55 (0.55)
Observations	8,922	5,575	3,687
R-squared	0.14	0.22	0.12

Source: Authors own calculations based on firms information. Regressions include age at base-line and gender as controls as well as month, year and region fixed-effects. Inference based on clustered standard errors (mall-level). ***: 1%, **: 5%, *: 10%

Table A4: The effect of the intervention on total number of transactions
Individual Level Analysis

Controls	Pooled (1)	Sales associates (2)	Managers (3)
D_i	-31.04 (20.38)	-1.44 (21.84)	1.75 (23.11)
B_t	-3.07 (10.85)	-3.83 (9.63)	-20.74 (19.25)
Parameter of interest ($D_i \times B_t$)	42.05** (19.29)	60.00** (25.10)	10.50 (21.60)
Observations	8,922	5,575	3,687
R-squared	0.11	0.16	0.13

Source: Authors own calculations based on firms information. Regressions include age at base-line and gender as controls as well as month, year and region fixed-effects. Inference based on clustered standard errors (mall-level). B Is defined so that the pre-treatment period includes information from 2014 and 2015 but only in the same months observed after treatment (September and October for sales associates and July to October for Managers) ***: 1%, **: 5%, *: 10%