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# MEASURING AND EXPLAINING MANAGEMENT IN SCHOOLS: NEW APPROACHES USING PUBLIC DATA 

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#### Abstract

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JEL Classification: M5, I2, J3
Keywords: Management, teacher selection, teacher incentives, Cross-country
Clare Leaver - clare.leaver@bsg.ox.ac.uk
University of Oxford, Blavatnik School of Government and CEPR
Renata Lemos - rlemos@worldbank.org
World Bank and CEP-LSE
Daniela Scur - daniela.scur@gmail.com
Cornell University, Dyson School of Applied Economics and Management and CEP-LSE and CEPR

# Measuring and explaining management in schools: New approaches using public data 

Clare Leaver<br>Blavatnik School of Government, University of Oxford<br>CEPR

Renata Lemos<br>World Bank<br>CEP-LSE

Daniela Scur<br>Dyson School of Applied<br>Economics and Management, Cornell University<br>CEPR, CEP-LSE

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#### Abstract

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## 1 Introduction

Despite global calls for improvements in education, progress towards learning for all is slow. This deficit is particularly pronounced for poor children and children in low-income countries [Akmal and Pritchett, 2019]. But why do some students learn more in some schools than others? While there are many contributing factors at system, school, and household-level, one consideration receiving growing attention is school management-the processes and practices used by principals day-today as they run their schools [World Bank, 2018]. Academics and practitioners interested in this issue face two challenges: how to measure school management accurately and cost-effectively at scale across schools and countries; and how to explain any observed relationship between school management and learning outcomes in a way that elucidates the underlying mechanisms to guide policy. This paper addresses both of these challenges.

Our first contribution is to develop a new approach to measurement that can, in principle, be used with any existing public dataset containing items about school management. We illustrate using two public datasets as examples: the OECD's Programme for International Student Assessment (PISA), and the Brazilian school census survey, Prova Brasil. The essence of our approach is to benchmark against the "state of the art", but expensive, World Management Survey (WMS) in Bloom et al. [2015a]. We show how questions from these public surveys can be classified into WMS topics (53 PISA questions into 14 WMS topics and 33 Prova Brasil questions into 8 WMS topics), how the responses can be coded using the WMS scoring rubric, and finally how these grades can be built into a school management index. Our PISA-based index covers over 15,000 schools across 65 countries, and our Prova Brasil-based index covers nearly all public schools in Brazil. These indices are well-validated and can be used by researchers interested in studying the role of management in education systems across a far wider range of countries and schools than was previously possible. ${ }^{1}$ All three indices, WMS, PISA, and Prova Brasil, show a strong, positive (within-country) correlation between school management and student learning outcomes, echoing recent causal evidence from randomized controlled trials in the U.S. [Fryer, 2014, 2017].

Our second contribution is to develop a framework to explore why management matters for schools. We set out, in general terms, how the impact of school management can be decomposed into learning gains that arise because given actors (teachers, students and parents) become more productive, and learning gains that arise because different actors join the school. To explore why these incentive and selection effects might arise, we turn to a specific model that captures key features of education systems in Latin America.

This model has two main building blocks. The first is the education production function: we assume that student learning depends on teacher ability, teacher effort, and household effort. The second is the impact of management practices where, considering the personnel policy restrictions the public sector faces, we distinguish between operations and people management. Good people management

[^0]practices enable managers to observe and contract on the performance of their employees, as well as to cultivate the intrinsic motivation of their staff. Good operations management practices enable managers to use resources efficiently and hence offer a higher level of teacher compensation and a more stimulating environment for students. ${ }^{2}$

Our framework predicts that good people management practices increase expected test scores through two channels. A teacher with a given ability and intrinsic motivation to teach exerts more effort because these practices provide extrinsic, and cultivate intrinsic, incentives. Compounding this, good people management practices improve selection: a teacher with high ability and high intrinsic motivation prefers a school with performance pay over alternative employments because she anticipates that she will work hard and be rewarded for producing student learning. We focus on Latin American countries and find support for both mechanisms in our PISA data. Principals in schools with higher PISA-based people management scores (predominantly private schools) are less likely to report experiencing teacher shortages and also report higher levels of teacher motivation and effort, compared to principals in schools with lower PISA-based people management scores.

Our framework also predicts that good operations management practices increase expected test scores through two channels. There is no teacher incentive effect but the selection effect remains, now driven by the level rather than structure of compensation. This is reinforced by a household incentive effect that arises because strong operations management practices encourage both students and parents to increase their inputs. We also find evidence of these mechanisms in our PISA data for Latin America. Principals in public schools with higher PISA-based operations management scores are less likely to report experiencing teacher shortages and also report higher levels of teacher motivation, teacher effort and household effort, compared to principals in public schools with lower PISA-based operations management scores.

While this is not definitive causal evidence, this combination of theory and descriptive empirical analysis offers a novel insight into why management matters in schools and we therefore move on to consider policy implications. People management practices such as performance pay, while common in the private sector, may not be possible in public schools. But there would seem to be fewer barriers to conducting assessments to judge teacher effectiveness, and letting such appraisals lead to changes in public recognition, opportunities for professional development, likelihood of career advancement, and/or greater responsibilities. That is, these people management practices help to attract, develop and reward good performers, and, our analysis suggests, should improve both teacher selection and incentives.

There is also substantial variation in the strength of operations management practices within the public sector. This suggest a role for government to encourage principals in public schools with weak operations management to follow best practices. Specific areas suggested by our analysis include

[^1]processes that facilitate: personalization of learning; dialogue among staff, students and parents focused on continuous improvement; and collection and use of student assessment data.

Related literature. Our first contribution - a new approach to measure management practices in schools - relates to two bodies of work. The first is the literature that has evolved since the creation of the WMS dataset first described in Bloom and Van Reenen [2007]. The WMS methodology has been adapted to a range of public sector institutions, including schools and universities [Bloom et al., 2015a, McCormack et al., 2014], healthcare facilities [Bloom et al., 2015b, 2019b], social programs [Delfgaauw et al., 2011, McConnell et al., 2009], and the civil service [Rasul and Rogger, 2016], as well as to low-income settings [Lemos and Scur, 2016]. However, it is expensive and timeconsuming to implement at scale; our approach is a feasible alternative. The second is the literature studying the role of education systems and institutions in determining student performance across countries [Wössmann, 2016]. Many recent papers use PISA data and have looked at this issue through the lens of autonomy [Hanushek et al., 2013, Wössmann et al., 2007], competition [West and Wössmann, 2010], student tracking [Hanushek and Wössmann, 2006, Ruhose and Schwerdt, 2016], external exams [Wössmann, 2005], and instructional time [Lavy, 2015]. Our PISA-based index enables researchers to consider school management in such studies.

Our second contribution - a framework to explain why management matters in schools-relates to the literature in personnel economics exploring incentives and selection. These channels have featured in prior work seeking to explain the performance of private sector employees [Bender et al., 2018, Cornwell et al., 2019, Lazear, 2000], public sector employees [Finan et al., 2017, Prendergast, 2007] and politicians [Besley, 2004, 2006, Gagliarducci and Nannicini, 2013, Martinez-Bravo, 2014]. Most closely related is Lazear [2003], who emphasises the potential selection margin of teacher performance pay, albeit without fully working up a formal model. ${ }^{3}$ A selection margin also features in the dynamic occupational model of Rothstein [2015] and the Roy model of Biasi [2019]. We study a wider range of management practices (beyond just performance pay) and provide an intuitive decomposition of the impact of these practices on student learning into incentives and selection.

The remainder of this paper is organized as follows. In Section 2, we set out our approach to measure management practices in schools, illustrating with the construction and validation of PISA-based and Prova Brasil-based management indices. In Section 3, we describe our theoretical framework, its testable predictions, a series of corroborative descriptive analyses from across Latin America, and the policy implications of these results. Section 4 concludes.

[^2]
## 2 How to measure management in schools?

Until the early 2000s, management was typically viewed as an unmeasurable productivity shifter, to be relegated to the residual in any performance regression [Bloom and Van Reenen, 2007]. Since then, improvements in survey methodology and data access have allowed for advances in measurement. The current "state of the art" approach uses a dedicated survey - the World Management Survey (WMS) - to measure establishments' adoption of structured management best practices. While the WMS offers uniquely rich information about management practices, it costs approximately USD400 per interview and takes about 4 months to conduct a single country wave [Bloom et al., 2016]. In view of these costs, it may not be well-suited to every context.

In this section, we propose an alternative three-step approach than can, in principle, be used with any existing public dataset containing information on management practices. The first step is to use the original WMS phone survey as a benchmark, and to look for questions in the public survey that elicit information on the management practices already measured by the WMS. ${ }^{4}$ The second step is to code answers in line with the WMS methodology. And the final step is to create a management index. In Section 2.1, we provide a brief overview of the WMS questions and coding. In Section 2.2, we describe our approach using two existing public datasets as examples: PISA and the Brazilian school census survey, Prova Brasil. Since Brazil and several other PISA countries are part of the Bloom et al. [2015a] sample, we can compare the (within-country) distribution of each index with the corresponding (within-country) distribution of the WMS index. Both indices are well-validated and can therefore be used by researchers interested in studying management across a wider range of countries and schools than was previously possible.

### 2.1 Overview of the World Management Survey methodology

The WMS was developed to measure adoption of structured management best practices in establishments across a range of countries and industries. ${ }^{5}$ The rigorous data collection is based on double-blind, semi-structured interviews conducted by highly-trained analysts and monitored by supervisors experienced on the survey methodology. Following its successful implementation in the private sector, the WMS was subsequently extended to public sector organizations [Bloom et al., 2015a, 2019b]; in this paper, we focus on the latter.

The public-sector WMS covers 20 topics across two main areas: operations management and people management. Broadly speaking, operations management in schools covers practices including: whether the school has standardization of instructional processes across classrooms while allowing

[^3]for within-classroom personalization of learning; whether and how the school uses assessments and data; and whether and how the school sets and uses targets and keeps track of progress. People management covers practices in handling good and bad performance measuring whether there is a systematic approach to identifying good and bad performance, rewarding school teachers proportionately, dealing with underperformers, and promoting and retaining good performers.

For each WMS topic, there is a scoring grid ranging from 1 to 5 , which serves as a guide to evaluate answers to questions during the semi-structured interviews. A score between 1 to 2 refers to a school with practically no structured management practices or very weak management practices implemented; a score between 2 to 3 refers to a school with some informal practices implemented, but these practices consist mostly of a reactive approach to managing the school; a score between 3 to 4 refers to a school where a good, formal management process is in place (though not yet consistent enough) and these practices consist mostly of a proactive approach to managing a school; and a score between 4 to 5 refers to well-defined, strong processes in place which are often seen as best practices in education. The overall management index, which measures the level of adoption of structured management best practices, is simply the average of the scores for these 20 topics.

The practices measured by the survey seem to matter: Bloom et al. [2015a] show that their school management score is strongly positively correlated with school-level student outcomes across 6 WMS countries (Brazil, Canada, India, Sweden, UK and US). ${ }^{6}$ They find a strong positive correlation for these countries: moving from the bottom to the top quartile of management is associated with a large increase in student learning outcomes, equivalent to approximately 0.4 standard deviations.

### 2.2 A new approach using existing public datasets

We now describe our approach, illustrating with the examples of PISA and Prova Brasil. ${ }^{7}$.

Construction. In 2012, alongside its famous student proficiency tests, PISA ran school principal surveys across 65 countries which included a wide-range of questions on both operations and people management. ${ }^{8}$ As a first step, we classified each of the PISA questions that could fall under one of the WMS topics, identifying 53 PISA questions that fit into 14 of the WMS topics. ${ }^{9}$ As a second step, we manually assigned scores for each of these PISA questions following the spirit of the scoring grid of the WMS and the US Census Management and Organizational Practices Survey (MOPS). As a final step, we built the overall management index, and the operations and people management subindices, following Anderson [2008]. This methodology weights the impact of the included variables

[^4]by the sum of their row in the inverse variance-covariance matrix, thereby assigning greater weight to questions that carry more "new information". ${ }^{10}$

PISA data is excellent for cross-country analysis, but it precludes in-depth analyses within countries as the sample of schools per country is typically small and does not include the necessary identifiers. Many countries, however, conduct their own national detailed surveys with school principals, teachers, and students in addition to administering standardized tests across grades. Latin America is particularly prolific: for example, Brazil's Prova Brasil, Colombia's SABER, Chile's SIMCE, and Peru's ECE are all available to researchers. These questionnaires provide rich information about practices at the school, as reported by a range of actors. In addition, the samples are usually large (often census-based) and contain school identifiers, thereby enabling researchers to explore heterogeneity and answer a wide range of policy-relevant questions. We illustrate how our approach can be applied widely to other national surveys using the example of Prova Brasil. This national survey plays a significant role in Brazil's education policy because its test results, along with promotion, dropout, and retention rates, are the main inputs to the Índice de Desenvolvimento da Educação Básica (IDEB), a national index representing educational quality at the school, municipality, and state levels.

We followed the same steps to create a Prova Brasil-based management index: we classified 33 questions (14 from the principal questionnaire and 19 from the teacher questionnaire) into 8 WMS topics, coded responses following the same rubric, and used the Anderson [2008] method to build a school management index.

Potential concerns. One of the key differences between the WMS survey and PISA or school census surveys is that the WMS is administered and analyzed by an independent interviewer, while the latter surveys are self-reported. There are a number of issues with self-reported data: for example, problems with translation and interpretation, and/or measurement equivalence. To address measurement error of cross-cultural understandings and norms on answering questions in our PISA index, we standardize our PISA-based management index within countries. This has an important implication: since all 65 countries have a mean score of zero, our index cannot be used to construct cross-country rankings of school management. Instead, the value of our PISA-based index lies in enabling academics and practitioners to study the (within-country) correlation between management and other variables for a far wider set of countries than was previously possible. ${ }^{11}$ This is not a concern for country-specific national surveys.

[^5]Another concern with self-reported data is that it is difficult to assess whether respondents are being accurate and truthful. In the WMS there are several strategies to elicit truthful information during the interview (such as always asking open-ended questions and asking for examples), but these are not available in self-reported questionnaires. We address this issue by focusing on the topics that have a direct equivalent in the WMS to allow for a clear benchmark for our new index. If principals are reporting "good" information in these surveys that allow us to capture similar signals as the WMS scores, we should see similar distributions of scores across the common countries and a similar overall relationship between management and student test scores across countries. For PISA, we compare the distribution of scores and the performance correlations for the common countries as there are no school identifiers available. For Prova Brasil, we use school identifiers to match schools directly and hence provide a one-to-one comparison of the index standardized values.

Validation of new indices. As a first validation exercise, we compare the distribution of our PISA-based management index with the distribution of school management as measured by the WMS data in Figure 1 for all countries that the WMS has collected data. ${ }^{12}$ The PISA and WMS distributions are reassuringly similar. The Kolmogorov-Smirnov test for equality of distributions rejects in only one of the 9 cases, Italy, where the PISA-index is somewhat more dispersed. ${ }^{13}$

As a second validation exercise, and to ensure we are picking up important variation with our management index, we conduct a basic check of the correlation between our measure and student performance. For each country we separate schools into quartiles of the management measure, and in Figure 2 we show, for each quartile, the average PISA test scores for math, reading and science (in deviations from the global mean). The graph includes all students and schools across the 65 countries available in the 2012 PISA dataset. This simple relationship suggests that students in schools in the bottom quartile of management within their country score are, on average, about 6 points lower than the PISA global mean, while students in schools in the top quartile of management within their country score, on average, about 5.5 points higher than the PISA global mean. To put this into context, 41 PISA points in math are the equivalent of a year of learning. The range of our results mirror how much, for example, the UK average science score changed between 2009 and 2015 (5 points), and how much the Brazilian average science score decreased over the same period (4 points).

Unlike PISA, the data in Prova Brasil includes school identifiers that allow for a one-to-one match with the schools surveyed in the 2013 WMS wave. ${ }^{14}$ In total, we have 262 matched schools in the public sector. We use this matched sample in Figure 3 where we show a school-level binned

[^6]scatter plot of WMS management score against the Prova Brasil-based management score. There is a positive and significant correlation of 0.19 , suggesting reasonable internal validation of the Prova Brasil index. As with the PISA index, we repeat the exercise of correlating the new index with student performance in secondary schools and find the same pattern (see Figure 4).

In Table 1 we formalize these relationships by reporting the average correlations between student learning and our management indices. For the student-level PISA dataset, we run OLS regressions via the OECD's repest Stata command, which uses the five available test score plausible values for each student and subject. We report the standard errors in parentheses and $p$-values in square brackets. The standard errors are clustered at the school level and use the appropriate survey weights. ${ }^{15}$ In the PISA specifications we include country fixed effects, and successively introduce school controls (dummies for school location, student-teacher ratio, log of the number of students, share of government funding relative to total funding the school receives, and ratio of computers connected to the web used as a proxy for school resources) and then student controls (gender, grade, socio-economic status and immigration status). All panels use the same sample but have different subject outcome variables. The R-squared for each set of regressions is reported within each panel, while the common sample characteristics and controls included are reported at the bottom of the table. Column (1) shows the raw relationship between the PISA-based school management index and student performance, only controlling for country fixed effects. The raw relationship ranges from just over 4 to almost 5 points on the PISA scale. ${ }^{16}$ Recall that 41 points on the PISA scale for math is equivalent to about one year of learning, and thus the raw correlation is equivalent to about one month of learning for math (similar for the other subjects). Column (2) includes school controls, which absorb little of the variation, and Column (3) shows the fully-specified regression including student controls. These controls account for a further point in the student performance. While we refrain from ranking management indices across countries, Figure 5 plots the coefficients of country-level regressions of management on PISA math test scores using the specification of Column (1). The estimation loses precision once we restrict to individual country samples but still broadly supports the positive relationship found in Table 1.

For the Prova Brasil student-level dataset, we run standard OLS regressions, also clustering the standard errors at the school level. Results are reported in standard deviations. In these Prova Brasil specifications, we include state fixed effects, and successively introduce school controls (studentteacher ratio, log of the number of students, dummy variables indicating the presence of an IT lab, science lab, and library, a dummy for male principals, dummies for educational attainment, and dummies for experience as principal), and then student controls (gender, race, socio-economic status, and mothers' educational attainment). Column (4) shows the raw relationship between the Prova Brasil-based school management index and student performance, only controlling for state fixed effects. One standard deviation higher score in the management index is strongly correlated with 0.068 standard deviations higher Portuguese scores and 0.078 standard deviations higher math

[^7]scores. Column (5) shows that including school characteristics absorbs very little of the variation, while Column (6) shows that including student characteristics absorbs only slightly more. The fully specified regression supports the general positive relationship between the school management index and student performance.

## 3 Why does management matter in schools?

There are myriad uses of our new indices. In this paper, we push the frontier of understanding the mechanisms behind the management and performance relationship by focusing on teachers. Our aim is not to provide a theoretical contribution per se, but rather to formalize a policy discussion around teacher incentive and selection mechanisms and their relationship to management practices and student performance. We take wider system-level factors-in particular hiring and firing autonomy, admissions autonomy and competition between schools-as given and assume that teachers and students make choices within the confines of this environment.

Real-world education systems are diverse, and in particular the dynamics of the public and private sector - and the type of private sector offerings - are different across countries. In some contexts, private schools target affluent households, and jobs in private schools are often seen as more attractive than public sector jobs, typically providing some form of performance-based compensation. In other contexts, there has been a growth of 'low-cost private schools' that deliberately cater for the lower end of the income distribution and, in these settings, jobs in the public sector typically confer significant rents relative to the private sector.

In view of this diversity, we focus our model and empirical test on one particular regional system: Latin America. We choose Latin America because its education systems are reasonably homogeneous across countries in terms of the character of public and private schools. Specifically, the private school system caters to the middle (and upper) classes and accounts for about one-fifth of high school students. Private schools tend to be better funded (via costly school fees) and in turn pay higher teacher salaries and offer better facilities. Public schools, on the other hand, are often poorly funded and operate in highly centralized environments. Teacher pay is set on a rigid scale dictated by strong unions. Focusing on a region that has such systems makes the applied theory exercise substantially less complex, and the prevalence of large-scale national surveys in the region opens many possibilities for future empirical work.

Table 2 reports the correlation exercise in Columns (1) to (3) of Table 1 for Latin American countries only, and confirms that the relationship between management and student performance is strong in the region. Further, the coefficient on private schools indicates that students in private schools achieve higher scores, by about 55 points, than students in public schools. This affords a suitable empirical environment to study the channels that we are interested in this section.

### 3.1 Overview of the theoretical framework

The analysis is built around a student-level education production function. A common, general formulation is $y=A(L, K)+\varepsilon$ where $y$ is a measure of student learning, $L$ and $K$ are respectively labour and physical capital inputs into the student's education, $A$ is a (school-specific) productivity parameter, and $\varepsilon$ is an error term. Here, we specialize to $y=\theta e+a+\varepsilon$, where $\theta$ is teacher ability, $e$ is teacher effort, and $a$ is household (student and/or parent) effort. That is, we enrich the specification of labour to allow for (additively separable) teacher and household inputs but abstract from the role of physical capital and school-level productivity. ${ }^{17}$ Using a theoretical framework built around this education production function, we show that school management structures can impact student learning outcomes via the following three channels:

1. Teacher selection: schools with high management scores offer compensation packages that select in more able (higher $\theta$ ) and more intrinsically motivated (lower effort cost) teacher types.
2. Teacher incentives: schools with high management scores offer compensation packages that extrinsically incentivize, and adopt practices that intrinsically motivate, more effort from any given teacher type that selects in.
3. Household incentives: schools with higher management scores institutionalize a strong work ethic and culture of high achievement among students and encourage greater parental involvement within the school (higher $a$ ).

In Section 3.2, we present a simple model that suffices to make the points above. In Section 3.3 and Section 3.4 we explain, intuitively, how the above selection and incentive effects are driven by people management and operations management. Then, in Section 3.5 we draw together these results and discuss implications for policy. We also briefly comment on how predictions would change in an alternative model featuring 'low-cost private schools'.

### 3.2 The model

We focus on a teacher who must decide whether to accept a job offer in her assigned public school, or decline it and apply to a private school or the outside sector.

Preferences. The teacher is risk neutral and cares about her compensation $w$ and effort $e$. When working in the education sector, the teacher's preferences are $w-\left(e^{2}-c e\right)$. The parameter $c$

[^8]captures her intrinsic motivation. This is because for $e<c / 2$ she derives a marginal benefit from exerting an extra unit of effort in teaching; it is only when $e>c / 2$ that effort costs kick in. We assume that $c=\tau+\Delta$. The first component $\tau$ denotes the teacher's baseline intrinsic motivation. This can be thought of as the realization of a random variable with density function $f$. The teacher observes this realization perfectly, while (at the time of hiring) employers observe nothing. The second component $\Delta$ is a motivational increment that, as we describe below, is determined by the people management practices in the teacher's chosen school. When working in the other sector, the teacher's preferences are simply $w-e^{2}$; intrinsic motivation plays no role. We abstract from differences within classes and focus on a representative household (student plus parents). This household cares only about its effort level $a$, and has preferences $-\left(a^{2}-\gamma a\right)$. The parameter $\gamma$ is a motivational increment that is also determined by management practices.

Performance metrics. Let $y_{1}$ denote a representative student's learning outcome in a school that hires the teacher, and $y_{0}$ denote a representative student's learning outcome in a school that does not hire the teacher. To the extent that teachers contribute to learning, one would expect $y_{1}>y_{0}$. We capture this in a simple way by assuming $y_{1}=\theta e+a+\varepsilon$ and $y_{0}=a+\varepsilon$. If the teacher is not hired by a school but instead chooses to work in the outside sector, her performance is $z=\theta e+\varepsilon$. The component $\theta$ denotes the teacher's ability. This can be thought of as the realization of a random variable with density function $g$, and which is drawn independently of $\tau$. The teacher observes this realization perfectly, while (at the time of hiring) employers observe nothing. Draws of the error term $\varepsilon$ are independent across employments. We assume throughout that $\varepsilon$ is mean zero and distributed $U[\underline{\varepsilon}, \bar{\varepsilon}]$. At times, for the purposes of illustration, we also assume a specific (uniform) distribution for $\theta$, as part of a numerical example that we discuss at the end of this section. ${ }^{18}$

Compensation schemes. Schools offer either a performance-pay contract or a fixed wage contract. Under the former, the teacher receives a base wage of $W$ plus a bonus $B$ if her performance exceeds a threshold $\bar{y}$. Under the latter, the teacher simply receives a base wage of $G$. The outside sector offers a performance-pay contract with a low base wage (normalized to zero) and a bonus $\beta$ if performance exceeds a threshold $\bar{z}$.

The impact of management practices. We assume that people management has two effects. The first relates to the structure of compensation: good people management practices enable managers to observe, and contract on, the performance of their employees-i.e. to offer a performancepay contract. The second relates to teacher motivation: good people management practices enable managers to cultivate the intrinsic motivation of their staff-i.e. to increase $\Delta$. We assume that

[^9]operations management also has two effects. The first relates to the level of compensation: good operations management practices free up resources and enable managers to offer a higher level of base pay. The second relates to household effort: good operations management practices help to create a stimulating environment for students and parents -i.e. to increase $\gamma$.

We classify schools into three management types: high (strong people and strong operations management), intermediate (weak people but strong operations management), and low (weak people and weak operations management). Performance metrics are indexed accordingly by $i=H, I, L$. We assume that high management schools are found exclusively in the private sector, while the public sector consists of a mix of intermediate and low management schools. This implies that performance-pay contracts are only offered by private schools (and the outside sector). Figure 6 provides evidence that this key assumption is well-supported in our PISA data. Here, we plot empirical cumulative distribution functions (CDFs) of the PISA-based people management score by sector and find that the private sector CDF (dashed blue plot) clearly first order stochastically dominates the public sector CDF (solid red plot). ${ }^{19}$

Timing. The timing of the game is as follows.

1. Nature chooses the teacher's two-dimensional type. This realization $(\tau, \theta)$ is observed by the teacher but not by employers.
2. Employers announce management structures and compensation schemes.
3. The teacher is assigned (by government) to a public school and decides whether to accept this post or decline it and apply either to a private school or the outside sector. ${ }^{20}$
4. Having made an occupational choice, the teacher chooses an effort level. Simultaneously, if the teacher is in the education sector, households choose effort levels.
5. A performance metric is realized. The teacher is rewarded in accordance with the compensation scheme announced at Stage 2.

Numerical example. At times in the analysis below, we will invoke specific distributional and parameter assumptions. In this numerical example, teacher intrinsic motivation is distributed $\tau \sim$ $U[0,10]$, and teacher ability is distributed $\theta \sim U[1,5]$. These random variables are independent of each other and the error term in the production functions. In a high management private school: teacher pay is $W+B=55$ if $y_{1}^{H} \geq 4.5$ and $W=15$ otherwise, and the motivational increments are $\Delta=0.5$ and $\gamma=2$. In an intermediate management public school teacher pay is $G=35$, and the

[^10]motivational increments are $\Delta=0$ and $\gamma=2$. And in a low management public school: teacher pay is $G=30$, and the motivational increments are $\Delta=0$ and $\gamma=1$. Pay in the outside sector is $\beta=50$ if $z \geq 1$ and 0 otherwise.

Our interest lies in establishing the impact of management practices on student learning via teacher occupational choice and effort level, and household effort level. We do not model the government's assignment rule, or the school principal's choice of management structure, simply treating these as exogenous parameters. The model is straightforward to solve (see Appendix A for details) and yields the insights summarized in the next two sections.

### 3.3 The impact of good people management

In this subsection, we use the theoretical framework to give a possible explanation for why schools with good people management may produce better student outcomes. In Section 3.3.1, we decompose the test score gain from people management into two effects: teacher selection and teacher incentives. If this decomposition is correct, then we should see evidence of these mechanisms in intermediate school outcomes. We develop this argument, and present corroborative evidence from our PISA dataset, in Section 3.3.2.

### 3.3.1 Decomposing the test score gain into teacher selection and incentives

A school of management type $i$ hires the teacher if, given her $(\tau, \theta)$ type, she expects to receive a higher payoff teaching in this school compared to other schools or working in the outside sector. Let the set of $(\tau, \theta)$ types hired by a school of management type $i$ be denoted by $\mathcal{T}^{i}$. The expected learning outcome of a representative student (i.e. ex ante, prior to occupational and effort choices) can therefore be written as

$$
\mathrm{E}\left[y^{i}\right]=\mathrm{E}\left[y_{1}^{i} \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{i}\right\}}\right]+\mathrm{E}\left[y_{0}^{i} \cdot 1_{\left\{\left[(\tau, \theta) \notin \mathcal{T}^{i}\right\}\right.}\right],
$$

where $1_{\left\{(\tau, \theta) \in \mathcal{T}^{i}\right\}}$ and $1_{\left\{\left[(\tau, \theta) \notin \mathcal{T}^{i}\right\}\right.}$ are indicator functions for the hiring and not hiring events. In keeping with the empirical application, we will refer to $\mathrm{E}\left[y^{i}\right]$ as the expected test score in school $i$.

The difference in expected test score across high and intermediate management schools-that is, the impact of people management holding operations management constant-can be written as

$$
\mathrm{E}\left[y^{H}\right]-\mathrm{E}\left[y^{I}\right]=\mathrm{E}\left[y_{1}^{H} \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}\right]-\mathrm{E}\left[y_{1}^{I} \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}\right],
$$

where the equality follows from the fact that people management only impacts test scores when the teacher is hired (the effect of household effort and the error term difference out). It is helpful to
decompose this difference as follows

$$
\begin{align*}
& \mathrm{E}\left[y^{H}\right]-\mathrm{E}\left[y^{I}\right]= \\
& \quad \underbrace{\mathrm{E}\left[\left(y_{1}^{H}-y_{1}^{I}\right) \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}\right]}_{\text {teacher incentives }}+\underbrace{\mathrm{E}\left[y_{1}^{I} \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}\right)\right]}_{\text {teacher selection }} . \tag{1}
\end{align*}
$$

The first term on the RHS of equation (1) captures what we will term the teacher incentive effect of good people management practices. Here, we compare the expected test score outcome in a high management private school with a teacher in the event that the teacher is hired to such a school against the expected test score outcome in an intermediate management public school with a teacher in the counterfactual event that the teacher is hired to a high management private school. In this way, we hold the set of ( $\tau, \theta$ ) types fixed and just consider how the incentive environment produces test scores.

In Lemma 1 in Appendix A, we derive teacher effort in high and intermediate management schools. Respectively, these are $e^{H}=\frac{\theta B}{2(\bar{\varepsilon}-\underline{\underline{\varepsilon}})}+\frac{\tau+\Delta}{2}$ and $e^{I}=\frac{\tau}{2}$. Substituting, we can write the first incentive term in (1) as

$$
\begin{align*}
& \mathrm{E}\left[\left(y_{1}^{H}-y_{1}^{I}\right) \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}\right]= \\
& \iint \theta(\frac{\theta}{2(\overbrace{B}^{\text {extrinsic }}}+\frac{\overbrace{\Delta}^{\text {intrinsic }}}{2}) \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}} f(\theta) g(\tau) d \theta d \tau . \tag{2}
\end{align*}
$$

We see from this expression that there are two teacher incentive channels. Part of the reason that test scores are higher in schools with good people management practices is because any given $(\tau, \theta)$ type exerts more effort due to: (i) an extrinsic incentive from the bonus $B$, and (ii) additional intrinsic motivation arising via the shift term $\Delta$.

The second term in equation (1) captures what we will term the teacher selection effect of good people management practices. Here, we compare the expected test score outcome in an intermediate management public school with a teacher in the event that the teacher is hired to such a school against the expected test score outcome in an intermediate management school with a teacher in the counterfactual event that the teacher is hired to a high management school. In this way, we hold the incentive environment fixed and just consider how the selection of $(\tau, \theta)$ types produces test scores. Substituting for $e^{I}$, we can write this second selection term as

$$
\begin{align*}
& \mathrm{E}\left[y^{I} \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}\right)\right]= \\
& \iint \overbrace{\theta}^{\text {ability }}(\frac{\overbrace{\tau}^{\text {effort }}}{2}) \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{H}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}\right) \quad f(\theta) g(\tau) d \theta d \tau . \tag{3}
\end{align*}
$$

We see from this expression that there are also two selection channels. A further part of the reason that test scores are higher in schools with good people management practices is because: (i) the $\tau$-types selected in are intrinsically motivated to exert more effort, and (ii) the $\theta$-types selected in are of greater ability.

To see this, consider the numerical example illustrated in Figure 7. In the top panel, the grey shaded area depicts the set of $(\tau, \theta)$ types that are hired by a high management private school. The unshaded area depicts the set of $(\tau, \theta)$ types that are hired by an intermediate management public school. It is clear that the intermediate management public school experiences negative selection on both dimensions. More able teachers prefer the performance-contingent compensation schemes available either in private schools or the outside sector. And more intrinsically motivated teachers prefer private schools because they anticipate exerting higher effort (and hence higher pay).

### 3.3.2 Predictions for intermediate school outcomes and evidence from PISA

Our theoretical framework suggests two mechanisms, teacher selection and teacher incentives, that could explain the positive correlation between people management scores and student learning outcomes apparent in the WMS, PISA and Prova Brasil data. If these mechanisms are correct, then we should see behavioural responses in intermediate school outcomes. In this section, we set out these predictions and then explore empirically whether they hold in our PISA data for Latin America. ${ }^{21}$

Teacher shortages. The probability of hiring the teacher in a high management private school is higher than the probability of hiring the teacher in an intermediate management public school (via teacher selection). In the numerical example shown in the top panel of Figure 7, the area of the grey region is bigger than the area of the unshaded region.

The PISA dataset does not contain objective information on school-level vacancies, so we use a series of 4 questions in the school principal questionnaire that ask the principal whether he/she feels that the school's capacity is hindered by a lack of qualified teachers in each of math, science, language and 'other subjects'. ${ }^{22}$ It is worth emphasising that these questions are open to considerable interpretation. For instance, a principal might answer 'a lot' because he/she feels that the school needs more new posts even if there are few vacancies for existing posts. Conversely, he/she might answer 'not all' because of a belief (or desire to say) that the school is coping despite there being

[^11]vacancies. ${ }^{23}$
With this caveat in mind, it can still be instructive to examine the data. Column (1) of Table 3 shows that, consistent with the theory, the teacher shortage index is 0.535 standard deviations lower among private schools than public schools, significant at the 1 percent level. Column (4) repeats the specification but with the people management index instead of the private school dummy. The relationship is consistently negative, suggesting that a one standard deviation increase in the people management index is correlated with 0.062 lower teacher shortage in schools, marginally significant at the 10 percent level.

Teacher motivation. The expected intrinsic motivation of a teacher hired to a high management private school is higher than the expected intrinsic motivation of a teacher hired to an intermediate management public school (via teacher selection and augmentation of teacher intrinsic motivation). In the numerical example in the top panel of Figure 7, the vertical height of the black point is greater than the vertical height of the blue point.

We explore this prediction by using the school climate section of the school principal questionnaire (questions relating to the perception of teachers' expectations of their students and meeting student needs, as well as the morale, enthusiasm, pride and valuation of academic achievement) to construct an index of teacher motivation. Column (2) of Table 3 shows that, consistent with the theory, the teacher motivation index is 0.591 standard deviations higher among private schools than public schools, significant at the 1 percent level. Using the people management index also yields a consistent relationship, shown in Column (5). The coefficient suggests a one standard deviation higher people management score is associated with 0.238 higher teacher motivation score, also significant at the 1 percent level.

Teacher effort. The expected effort level of a teacher hired to a high management private school is higher than the expected effort level of a teacher hired to an intermediate management public school (via teacher selection, extrinsic teacher incentives, and augmentation of teacher intrinsic motivation on-the-job)

$$
\mathrm{E}\left[\left.\frac{\theta B}{2(\bar{\varepsilon}-\underline{\varepsilon})}+\frac{\tau+\Delta}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{H}\right]>\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{I}\right] .
$$

We explore this prediction by using the school climate section of the school principal questionnaire (questions relating to how often teachers are absent, late and/or unprepared) to construct an index of teacher effort. Column (3) of Table 3 shows that, consistent with the theory, the teacher effort index is 0.792 standard deviations higher among private schools than public schools, significant at the 1 percent level. Column (6) reports the same specification for the people management index,

[^12]suggesting that a one standard deviation increase in the management score is associated with 0.074 higher teacher effort, significant at the 5 percent level.. For completeness, Figure 8 plots the coefficients for country-level regressions using the same specifications reported in Columns (4) to (6) of Table 3. While there is some variation across countries, the results broadly hold.

### 3.4 The impact of good operations management

In this subsection, we use the theoretical framework to give a possible explanation for why schools with good operations management may produce better student outcomes. In Section 3.4.1 we decompose the test score gain from operations management into three effects: teacher selection, teacher incentives and household incentives. If this decomposition is correct, then we should see evidence of these mechanisms in intermediate school outcomes. We develop this argument, and present corroborative evidence from our PISA dataset, in Section 3.4.2.

### 3.4.1 Decomposing the test score gain into teacher selection, teacher incentives, and household incentives

The difference in expected test scores across intermediate and low management public schools - that is, the impact of operations management holding people management constant-is

$$
\begin{aligned}
\mathrm{E}\left[y^{I}\right]- & \mathrm{E}\left[y^{L}\right]= \\
& \mathrm{E}\left[y_{1}^{I} \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}\right]-\mathrm{E}\left[y_{1}^{L} \cdot 1_{\left.(\tau, \theta) \in \mathcal{T}^{L}\right\}}\right]+\mathrm{E}\left[y_{0}^{I} \cdot 1_{\left\{(\tau, \theta) \notin \mathcal{T}^{I}\right\}}\right]-\mathrm{E}\left[y_{0}^{L} \cdot 1_{\left\{(\tau, \theta) \in \mathcal{T}^{L}\right\}}\right] .
\end{aligned}
$$

Letting $a^{I}$ and $a^{L}$ respectively denote household effort in these schools, and using the same decomposition as before, we can rewrite this difference as

$$
\begin{equation*}
\mathrm{E}\left[y^{I}\right]-\mathrm{E}\left[y^{L}\right]=\underbrace{\mathrm{E}\left[y_{1}^{L} \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{L}\right\}}\right)\right]}_{\text {teacher selection }}+\underbrace{a^{I}-a^{L}}_{\text {household incentives }} . \tag{4}
\end{equation*}
$$

There is no teacher incentive term because both extrinsic teacher incentives and augmentation of teacher intrinsic motivation depend on people management and this is assumed to be constant across these schools. We can write the teacher selection term as:

$$
\begin{align*}
& \mathrm{E}\left[y^{L} \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{L}\right\}}\right)\right]= \\
& \iint \overbrace{\theta}^{\text {ability }}(\frac{\overbrace{\tau}^{2}}{\text { effort }^{\mathcal{T}}}) \cdot\left(1_{\left\{(\tau, \theta) \in \mathcal{T}^{I}\right\}}-1_{\left\{(\tau, \theta) \in \mathcal{T}^{L}\right\}}\right) \quad f(\theta) g(\tau) d \theta d \tau \text {. } \tag{5}
\end{align*}
$$

Again, there are two teacher selection channels. Part of the reason that test scores are higher in schools with good operations management practices is because: (i) the $\tau$-types selected in are in-
trinsically motivated to exert more effort, and (ii) the $\theta$-types selected in are of greater ability.
To see this, consider the numerical example illustrated in Figure 7. The unshaded area in the top panel depicts the set of $(\tau, \theta)$ types that are hired by an intermediate management public school, while the unshaded area in the bottom panel depicts the set of $(\tau, \theta)$ types that are hired by a low management public school. It is clear that the intermediate management public school hires both more and better types.

In contrast to the people management case, there is a channel operating via household incentives. A further part of the reason that test scores are higher in schools with good operations management practices is because households (students plus parents) exert more effort due to additional intrinsic motivation arising via the shift term $\gamma$.

### 3.4.2 Predictions for intermediate school outcomes and evidence from PISA

Again, we set out predictions relating to intermediate school outcomes (see Remark 2 in Appendix A), and then explore empirically whether they hold in our PISA data. ${ }^{24}$

Teacher shortages. The probability of hiring the teacher in an intermediate management public school is higher than the probability of hiring the teacher in a low management public school (via teacher selection). In the numerical example shown in Figure 7, the unshaded area is larger in the top panel relative to the bottom panel.

We take this prediction to the data using the 4 questions in the PISA school principal questionnaire that ask the principal whether they feel that the school's capacity is hindered by a lack of qualified teachers (see Appendix C). Column (1) of Table 4 shows a negative and statistically significant correlation between the operations management index and the teacher shortage index. A one standard deviation increase in operation score is associated with a 0.076 standard deviation decrease in the teacher shortage index, significant at the 10 percent level.

Teacher motivation. The expected intrinsic motivation of a teacher hired to an intermediate management public school is higher than the expected intrinsic motivation of a teacher hired to a low management public school (via teacher selection). In the numerical example shown in Figure 7, the vertical height of the blue point in the top panel is greater than the vertical height of the orange point in the bottom panel.

Column (2) of Table 4 shows that, consistent with the theory, the partial effect of operations score on the teacher motivation index is positive and significant at 1 percent; a one standard deviation

[^13]increase in operation score is associated with a 0.238 standard deviation increase in the teacher motivation index, significant at the 1 percent level.

Teacher effort. The expected effort level of a teacher hired to an intermediate management public school is higher than the expected effort level of a teacher hired to a low management public school (via teacher selection)

$$
\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{I}\right]>\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{L}\right]
$$

Column (3) of Table 4 shows that one standard deviation higher operations score is correlated with 0.076 standard deviations higher teacher effort, significant at the 5 percent level.

Household effort The expected level of household effort in an intermediate management public school is higher than the expected level of household effort in a low management public school (via augmentation of student intrinsic motivation): $a^{I}>a^{L}$.

We explore this prediction by using the school climate section of the school principal questionnaire to construct an index of household effort, combining student behavior questions (relating to how often students are truant, late, disrespectful and/or disruptive) and parental involvement questions (relating to the extent to which parents: are interested in, and discuss, their child's progress and behaviour; volunteer for school activities; and participate in school governance or other forms of accountability). Column (4) of Table 4 shows that, consistent with the theory, one standard deviation higher operations management score is correlated with 0.160 higher household effort, significant at the 1 percent level. For completeness, Figure 9 plots the coefficients for country-level regressions using the same specifications in Columns (1) to (4) in Table 4. Again, the results broadly hold at country-level.

### 3.5 Summary of theoretical prediction

We developed a simple theoretical framework, built around a student-level education production function, to explore why management practices might matter in schools. Using this framework, we showed that people management practices may be contributing to higher student test scores through two channels: teacher selection and teacher incentives. The predictions that these channels imply for intermediate school outcomes-fewer teacher shortages, higher teacher motivation, and higher teacher effort in schools with strong people management than in schools with weak people management-are all well-supported in our PISA data for Latin America.

We also showed that operations management practices may be contributing to higher student test scores via two channels: teacher selection and household incentives. The empirical support for the predictions that these channels imply for intermediate school outcomes-fewer teacher shortages,
higher teacher motivation, higher teacher effort, and higher household effort in schools with strong operations management than in schools with weak operations management-is also strong.

While this does not represent definitive causal evidence, this combination of theory and descriptive empirical analysis offers an insight into why management appears to matter in schools and so we cautiously move on to policy. For example, what type of school management practices might be changed to drive improved student learning? Our analysis suggests that people management is a good place to start. While it may not be feasible (on political or budgetary grounds) for governments to introduce performance pay in public schools, it may be possible to conduct assessments to judge teacher effectiveness, and for these appraisals to lead to changes in public recognition, and to opportunities for professional development, likelihood of career advancement, and greater responsibilities and leadership. Such practices also reward and develop good performers and create a good employee proposition and could improve both teacher selection and incentives.

Beyond the difference in people management across private and public sectors, a striking feature of the PISA data is that there is substantial variation in the strength of operations management practices within the public sector. Some public schools are adopting management practices that appear to be driving student learning, while others within the same public education system are not. This suggests a role for government to encourage schools with weak operations management to follow best practice. As we observe from our mapping exercise, 'strong' operations management practices do two things, they: actively promote quality of delivery in the classroom (e.g. via personalization of learning, and encouragement to follow best educational practice); and put processes in place to review school performance and drive change (e.g. dialogue and meetings focused on continuous improvement, collection and use of student assessment data). Such practices could be adopted more widely in the public sector and, our analysis suggests, should improve both teacher selection and household incentives.

To be sure, our goal has not been to produce a global theory; given the diversity of real-world education systems, we have deliberately focused on one region, Latin America, and developed a theoretical framework for that context. One could adapt this framework to study different settings, for instance South Asia and parts of East Africa where there is a preponderance of 'low-cost private schools'. Such a model would need to allow for the possibility of 'queues' for jobs in public schools and, as a result, to explicitly model demand-side selection. To the extent that strong management practices enable public school principals to offer higher levels of compensation and then choose motivated and able teachers from the resulting queue, then qualitatively similar predictions would likely still apply. We hope that our new index will enable fertile ground for further research.

## 4 Conclusion

Policy makers have begun to set ambitious, universal learning goals. To achieve these targets it will be necessary to understand why, within and across current education systems, some students are learning more in some schools than others. Although there are likely many factors at work, it has been suggested that part of this variation in learning might stem from differences in school management. To explore this issue and develop policy, academics and practitioners need to be able to measure school management accurately and cost-effectively at scale across schools and countries, and be in a position to postulate mechanisms behind any observed relationship between school management and student learning outcomes.

This paper has responded to these observations by developing new approaches to measurement, as well as a simple theoretical framework that captures key features of education systems in Latin America. The first application of our new measurement approach used publicly available data from the school principal surveys conducted by PISA to construct a school management index spanning 65 countries. This PISA-based school management index can be well-validated against the more detailed (though also much more expensive) index based on the WMS. As such, it has clear value in settings where cross-country coverage is important, enabling researchers to study and compare the (within-country) correlation between management and student/school-level outcomes for a far wider set of countries than was previously possible.

Our second application used publicly available data from a national administrative survey to construct a school management index spanning all public schools in a single country: Brazil. This Prova Brasil-based index was also well-validated against the WMS. This second application has value in settings where within-country coverage, and the availability to merge with other administrative data sets, is important.

It is striking that both of our new school management indices confirm the strong positive correlation of school management scores with school-level student outcomes first reported in Bloom et al. [2015a]. A positive relationship holds for the global PISA sample, and in the census of schools in Brazil. Our theoretical framework, for the first time, formalizes the possible causal mechanisms in one of these regions: Latin America. We argued that strong people management practices may be improving student learning through a combination of teacher selection and incentive effects, and that schools could be encouraged to adopt practices that reward good performers, develop good performers, and create a good employee value proposition. Looking to operations management, we argued that strong operations practices may be improving student learning through a combination of teacher selection and household incentives, and that schools could be encouraged to adopt practices promoting quality of delivery in the classroom and adopt processes to review school performance and drive change. We also provided a suggestive set of evidence for these channels.

Improvements to management practices present an untapped opportunity for potentially large improvements in educational outcomes, particularly in cash-strapped regions of the world. One possible
way of effecting change is to support existing school principals to introduce stronger people and operations management practices, for instance via training and resources. Fryer [2014, 2017] reports positive results from RCTs injecting best management practices into U.S. public schools. Another possibility is to contract new managers into existing public schools. Romero et al. [forthcoming] report mixed results from an RCT in Liberia in which (non-governmental) management teams were contracted to run public schools: contracting-in raised learning outcomes, but new managers spent more and may have engaged in strategic behaviour. Investigating how to implement strong people and operations management practices to drive learning for all is an important area for future research.

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Table 1: School management and student performance

| Panel A: | PISA |  |  | Prova Brasil |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | (1) | (2) | (3) | (4) | (5) | (6) |
|  | Reading PISA points |  |  | Portuguese scores (SDs) |  |  |
| Management Index | $\begin{gathered} 4.904 \\ (1.193) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} \hline 3.947 \\ (1.172) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} \hline 3.019 \\ (0.980) \\ {[0.002]} \end{gathered}$ | $\begin{gathered} 0.068 \\ (0.003) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} 0.066 \\ (0.003) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} 0.059 \\ (0.002) \\ {[0.000]} \end{gathered}$ |
| Private |  | 11.514 (2.889) [0.000] | $\begin{gathered} 2.911 \\ (2.560) \\ {[0.255]} \end{gathered}$ |  |  |  |
| $R$-squared | 0.24 | 0.29 | 0.42 | 0.03 | 0.04 | 0.10 |
| Panel B: | Math PISA points |  |  | Math scores (SDs) |  |  |
| Management Index | $\begin{gathered} \hline 4.689 \\ (1.267) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} 3.937 \\ (1.272) \\ {[0.001]} \end{gathered}$ | $\begin{gathered} \hline 2.800 \\ (1.060) \\ {[0.008]} \end{gathered}$ | $\begin{gathered} 0.078 \\ (0.003) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} \hline 0.075 \\ (0.003) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} 0.068 \\ (0.002) \\ {[0.000]} \end{gathered}$ |
| Private |  | $\begin{aligned} & 11.467 \\ & (2.874) \\ & {[0.000]} \end{aligned}$ | $\begin{gathered} 2.001 \\ (2.655) \\ {[0.451]} \end{gathered}$ |  |  |  |
| $R$-squared | 0.31 | 0.34 | 0.45 | 0.05 | 0.05 | 0.10 |
| Panel C: | Science PISA points |  |  | n.a. |  |  |
| Management Index | $\begin{gathered} \hline 4.283 \\ (1.187) \\ {[0.000]} \end{gathered}$ | $\begin{gathered} \hline 3.601 \\ (1.217) \\ {[0.003]} \\ \hline \end{gathered}$ | 2.553 <br> (0.982) <br> [0.009] |  |  |  |
| Private |  | $\begin{aligned} & 10.215 \\ & (2.751) \\ & {[0.000]} \end{aligned}$ | $\begin{gathered} 1.245 \\ (2.377) \\ {[0.600]} \end{gathered}$ |  |  |  |
| $R$-squared | 0.30 | 0.33 | 0.43 |  |  |  |
| \# Observations | 410701 | 410701 | 410701 | 9890704 | 9890704 | 9890704 |
| \# Schools | 15196 | 15196 | 15196 | 33148 | 33148 | 33148 |
| Location FE* | Y | Y | Y | Y | Y | Y |
| School controls |  | Y | Y |  | Y | Y |
| Student controls |  |  | Y |  |  | Y |

Notes: Standard errors in parentheses, p-values in square brackets. OLS regressions for PISA were run with the student-level PISA dataset using the OECD's repest Stata command. Standard errors clustered at the school level and use all 5 plausible values for each subject and student final weights. Prova Brasil regressions run with standard OLS. Standard errors clustered at the school level and dependent variables are student learning outcomes on national tests at Grade 9 (the same exercise can be done with primary schools and tests at Grade 5). All specifications include location fixed effects (countries for PISA and states for Prova Brasil). PISA controls: School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, and ratio of computers connected to the web as a proxy for school resources. Student controls include gender, grade, socio-economic status and immigration status. Prova Brasil controls: School controls include student-teacher ratio, log of the number of students, dummies indicating the presence of an IT lab, science lab, and library as proxies for school resources. Given availability of principal characteristics, school controls also include a dummy for male principals, dummies for educational attainment, and dummies for experience as principal. Student controls include a dummy for male students, a dummy for white students, student households' consumption index, dummies for mother educational attainment (grades 1-5, grades 6-9, secondary grades 10-12, and college). For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value. All panels use the same sample but have different subject outcome yariables. Summary statistics for PISA dependent variables and controls are presented in Table B.1.

Table 2: PISA management index and student performance: Latin America

|  | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) | (9) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Reading PISA points |  |  | Math PISA points |  |  | Science PISA points |  |  |
| Management Index | 8.255 | 2.681 | 2.212 | 7.442 | 2.432 | 1.764 | 7.859 | 3.092 | 2.509 |
|  | (1.610) | (1.252) | (1.008) | (1.576) | (1.230) | (1.039) | (1.421) | (1.144) | (0.973) |
|  | [0.000] | [0.032] | [0.028] | [0.000] | [0.048] | [0.089] | [0.000] | [0.006] | [0.009] |
| Private |  | 56.807 | 31.921 |  | 55.695 | 32.589 | 0.000 | 55.428 | 33.077 |
|  |  | (3.301) | (2.956) |  | (3.713) | (3.121) |  | (3.735) | (3.327) |
|  |  | [0.000] | [0.000] |  | [0.000] | [0.000] |  | [8.161] | [2.736] |
| $R$-squared | 0.032 | 0.173 | 0.342 | 0.041 | 0.185 | 0.350 | 0.040 | 0.172 | 0.312 |
| \# Observations | 78144 | 78144 | 78144 | 78144 | 78144 | 78144 | 78144 | 78144 | 78144 |
| \# Schools | 3075 | 3075 | 3075 | 3075 | 3075 | 3075 | 3075 | 3075 | 3075 |
| \# Countries | 8 | 8 | 8 | 8 | 8 | 8 | 8 | 8 | 8 |
| Country FE | Y | Y | Y | Y | Y | Y | Y | Y | Y |
| School controls |  | Y | Y |  | Y | Y |  | Y | Y |
| Student controls |  |  | Y |  |  | Y |  |  | Y |

Notes: Standard errors in parentheses, p-values in square brackets. OLS regressions for PISA were run with the student-level PISA dataset for 8 Latin American countries using the OECD's repest Stata command. Standard errors clustered at the school level and use all 5 plausible values for each subject and student final weights. All specifications include country fixed effects. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, and ratio of computers connected to the web as a proxy for school resources. Student controls include gender, grade, socio-economic status and immigration status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values.

Table 3: People management and intermediate outcomes, public and private schools in Latin America

|  | $(1)$ <br> z-teacher <br> shortage | $(2)$ <br> z-teacher <br> motivation | $(3)$ <br> z-teacher <br> effort | $(4)$ <br> z-teacher <br> shortage | $(5)$ <br> z-teacher <br> motivation | $(6)$ <br> z-teacher <br> effort |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Private School | -0.535 | 0.591 | 0.792 |  |  |  |
|  | $(0.122)$ | $(0.139)$ | $(0.128)$ |  |  |  |
| People Index | $[0.000]$ | $[0.000]$ | $[0.000]$ |  |  |  |
|  |  |  |  | -0.062 | 0.238 | 0.074 |
| R-squared | 0.152 | 0.142 | 0.154 | 0.139 | $0.045)$ | $(0.040)$ |
| Observations | 3035 | 3043 | 3043 | 3035 | 3043 | 3043 |
| School controls | $Y$ | $Y$ | $Y$ | $Y$ | $Y$ | Y |
| Country FE | $Y$ | $Y$ | $Y$ | $Y$ | $Y$ | Y |

Notes: The first row reports the coefficients from regressions of a binary indicator (coded to 1 if the school is a private school, 0 otherwise) on the standardized index of three intermediate school outcomes: teacher shortage, teacher motivation and teacher effort. The second row reports coefficients from regressions of the standardized people management index on each of the intermediate school outcomes. The people management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include PISA school final weights and country fixed effects. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

Table 4: Operations management and intermediate outcomes, public schools in Latin America

|  | $(1)$ <br> z-teacher <br> shortage | $(2)$ <br> z-teacher <br> motivation | $(3)$ <br> z-teacher <br> effort | $(4)$ <br> z-household <br> effort |
| :--- | :---: | :---: | :---: | :---: |
| Operations Management Index | -0.080 | 0.238 | 0.076 | 0.160 |
|  | $(0.043)$ | $(0.041)$ | $(0.038)$ | $(0.054)$ |
| $R$-squared | $[0.061]$ | $[0.000]$ | $[0.044]$ | $[0.003]$ |
| Observations | 0.0787 | 0.171 | 0.154 | 0.242 |
| School controls | 2407 | 2414 | 2414 | 2414 |
| Country FE | Y | Y | Y | Y |

Notes: All regressions use data from public schools only. The table reports coefficients from regressions of the standardized operations management index on each of the intermediate school outcome. The operations management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include PISA school final weights and country fixed effects. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

Figure 1: Distribution of overall management scores, PISA vs WMS


Note: Data for the World Management Survey index for all countries except for Mexico and Colombia can be found at www.worldmanagementsurvey.org. Distribution of overall management indices standardized within countries. Kernel density curves estimated using WMS sampling weights (calculated as the inverse probability of being interview on log of number of students, public status, and population density by state, province, or NUTS 2 region as a measure of location) for the WMS data and school final weights for the PISA data. Samples include both public and private secondary schools for both datasets, with the exception of Colombia where WMS data is only available to public primary schools. Number of WMS/PISA observations are as follow (WMS $/$ PISA ) : Brazil $=510 / 561$, Canada $=129 / 770$, Colombia $=468 / 268$, Great Britain $=89 / 422$, Germany $=$ $102 / 158$, Italy $=284 / 926$, Mexico $=157 / 1327$, Sweden $=85 / 179$, United States $=263 / 136$.

Figure 2: PISA-based management index by quartile x PISA student outcomes


[^14]Figure 3: Prova Brasil-based management index x WMS management index


Note: This graph is a binned scatter plot. Each circle represents the average of 5 schools. The sample contains 262 schools which have data for both Prova Brasil and WMS in 2013. Correlation of 0.19 (p-level:0.00).

Figure 4: Prova Brasil-based management index by quartile x student outcomes


Note: The sample contains 33,148 public secondary schools of Prova Brasil in 2013 for which have available data. For simplicity and to compare the results to the results of PISA and the WMS, we use student learning outcomes on national tests in Portuguese and Math at Grade 9. The same exercise can be repeated with primary schools and national tests in Portuguese and Math at Grade 5.

Figure 5: Coefficient plot of PISA-based management index x math PISA points, by country


Note: PISA 2012 data. Regressions are estimated using OECD's repest command in Stata, by country. The specification includes all five plausible values for PISA 2012 and student final weights. Each marker represents the coefficient (and vertical spike represents associated $95 \%$ confidence intervals) of the management index on math scores.

Figure 6: Cumulative distribution of people management, Latin America


Notes: Cumulative distribution of the PISA-based people management index for private and public schools for 8 Latin American countries. The people management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. Sample consists of 3075 schools: 2432 in the public sector and 637 in the private sector.

Figure 7: Teacher selection


Note: The blue point in the top panel shows average teacher ability $\theta$ and baseline intrinsic motivation $\tau$ among teacher types who select into an intermediate management public school; the black point in the same panel shows average $\theta$ and $\tau$ among teacher types who select into a competing high management private school. The orange point in the bottom panel shows average $\theta$ and $\tau$ among teacher types who select into a low management public school; the black point in the same panel shows average $\theta$ and $\tau$ among teacher types who select into a competing high management private school. Both panels are plotted for the numerical example set out in Section 3.2.

Figure 8: Coefficient plot of PISA-based people management index x intermediate outcomes, by country


Notes: Each marker represents the coefficient (and vertical spike represents the associated $95 \%$ confidence intervals) from regressions of the people management index on the intermediate school outcome indices for each country in Latin America. The people management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include country fixed effects, school controls and PISA school final weights. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

Figure 9: Coefficient plot of PISA-based operations management index x intermediate outcomes, by country



Notes: Each marker represents the coefficient (and vertical spike represents the associated $95 \%$ confidence intervals) from regressions of the operations management index on the intermediate school outcome indices for each country in Latin America. The operations management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include country fixed effects, school controls and PISA school final weights. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

## A Appendix: Theoretical derivations

Lemma 1. Assume that the government assigns the teacher to an intermediate management public school.

1. If the teacher accepts the government's offer, then she exerts effort $e^{I}=\frac{\tau}{2}$.
2. If the teacher declines the government's offer and is hired by a high management private school, then she exerts effort $e^{H}=\frac{\theta B}{2(\bar{\varepsilon}-\underline{\varepsilon})}+\frac{\tau+\Delta}{2}$.
3. If the teacher declines the government's offer and is hired by an outside employer, then she exerts effort $e^{O}=\frac{\theta \beta}{2(\bar{\varepsilon}-\underline{\varepsilon})}$.

Proof. Part 1. When working in an intermediate management public school, a teacher with baseline motivation $\tau$ chooses effort to solve

$$
\max _{e} G-\left(e^{2}-(\tau) \cdot e\right)
$$

Differentiation to obtain the first order condition yields the solution stated above. (Here, as in the cases below, the second order condition necessary for a maximum holds.)

Part 2. When working in a high management private school, a teacher with baseline motivation $\tau$ and ability $\theta$ chooses effort to solve

$$
\max _{e} P \cdot B+W-\left(e^{2}-(\tau+\Delta) \cdot e\right)
$$

where $P$ is the probability that $y_{1}^{H}$ exceeds the threshold $\bar{y}$ given $e$ (and student attention a). Given the uniform distribution for $\varepsilon$, we can rewrite this probability as

$$
P=\operatorname{Pr}(\theta e+a+\varepsilon>\bar{y})=\operatorname{Pr}(\theta e+a-\bar{y}>-\varepsilon)=\frac{\bar{\varepsilon}+\theta e+a-\overline{\bar{y}}}{\bar{\varepsilon}-\underline{\varepsilon}} .
$$

The first order condition for this optimization problem is

$$
\frac{\theta B}{\bar{\varepsilon}-\underline{\varepsilon}}=2 e-(\tau+\Delta)
$$

which yields the solution stated above.
Part 3. When working in the outside sector, a teacher chooses effort to solve

$$
\max _{e} P^{O} \cdot \beta-e^{2}
$$

where $P^{O}$ is the probability that $z$ exceeds the threshold $\bar{z}$ given $e$. We can rewrite this probability
as

$$
P^{O}=\operatorname{Pr}\left(\theta e+\varepsilon^{O}>\bar{z}\right)=\operatorname{Pr}\left(\theta e-\bar{z}>-\varepsilon^{O}\right)=\frac{\bar{\varepsilon}+\theta e-\bar{z}}{\bar{\varepsilon}-\underline{\varepsilon}} .
$$

The first order condition for this optimization problem is

$$
\frac{\theta \beta}{\bar{\varepsilon}-\underline{\varepsilon}}=2 e,
$$

which yields the solution stated above.
Lemma 2. Assume that the government assigns the teacher to an intermediate management public school. There exist functions

$$
\tau^{G}=\frac{56}{8 \theta+1}-2 \theta-\frac{1}{4}, \quad \tau^{O}=\sqrt{25 \theta^{2}-60}, \quad \text { and } \quad \tau^{P}=\sqrt{25 \theta^{2}-4}-4 \theta-\frac{1}{2}
$$

such that:

1. The teacher accepts the government's offer with probability $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{I}\right]$, where $\mathcal{T}^{I} \equiv(\tau, \theta)$ : $\tau^{O}(\theta) \leq \tau \leq \tau^{G}(\theta)$.
2. The teacher declines the government's offer and accepts an offer from a private school with probability $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{H}\right]$, where $\mathcal{T}^{H} \equiv(\tau, \theta): \tau \geq \max \left\{\tau^{G}(\theta), \tau^{P}(\theta)\right\}$.

Proof. Part 1. The function $\tau^{G}$ traces out the loci of $(\tau, \theta)$-types who, anticipating subsequent teacher effort and household effort levels, are indifferent between accepting their government job offer and declining it in favour of a job in a high management private school, i.e. types for whom

$$
G-\left(e^{L}\right)^{2}+\tau e^{L}=B\left(\frac{\bar{\varepsilon}+\theta e^{H}+a^{H}-\bar{y}}{\bar{\varepsilon}-\underline{\varepsilon}}\right)-\left(e^{H}\right)^{2}+(\tau+\Delta) e^{H} .
$$

Substituting for $e^{L}$ and $e^{H}$ from Lemma 1, together with the parameters in our numerical example (implying $a^{H}=1$ ), and rearranging yields

$$
\tau^{G}=\frac{56}{8 \theta+1}-2 \theta-\frac{1}{4}
$$

Fixing $\theta$, for any $\tau<\tau^{G}(\theta)$, the teacher's payoff from accepting the government's offer is strictly higher than her expected payoff from declining and accepting a job in a high management private school.

The function $\tau^{O}$ traces out the loci of $(\tau, \theta)$-types who, anticipating subsequent teacher effort levels, are indifferent between accepting their government job offer and declining it in favour of a job in the outside sector, i.e. types for whom

$$
G-\left(e^{L}\right)^{2}+(\tau) e^{L}=\beta\left(\frac{\bar{\varepsilon}+\theta e^{O}-\bar{z}}{\bar{\varepsilon}-\underline{\varepsilon}}\right)-\left(e^{O}\right)^{2} .
$$

Substituting for $e^{L}$ and $e^{O}$ from Lemma 1, together with the parameters in our numerical example, and rearranging for $\tau$ yields

$$
\tau^{O}=\sqrt{25 \theta^{2}-60}
$$

Fixing $\theta$, for any $\tau>\tau^{O}(\theta)$, the teacher's payoff from accepting the government's offer is strictly higher than her expected payoff from declining and accepting a job in the outside sector.

All that remains, is to confirm that there exist values of $\theta$ such that $\tau^{O} \leq \tau^{G}$. Clearly, $\tau^{G}$ is decreasing and $\tau^{O}$ is increasing. Straightforward calculations show that $\tau^{G}-\tau^{O}$ is positive and decreasing on $[1,1.56]$ which establishes that there exists a set $\mathcal{T}^{I} \equiv(\tau, \theta): \tau^{O} \leq \tau \leq \tau^{G}$. For any pair $(\tau, \theta)$ in this set, the payoff from accepting the government job (weakly) exceeds both the expected payoff of declining and accepting a job in a high management private school and the expected payoff of declining and accepting a job in the outside sector.

Part 2. The function $\tau^{P}$ traces out the loci of $(\tau, \theta)$-types who, anticipating subsequent teacher effort and household effort levels, and having declined their government job offer, are indifferent between a job in a high management private school and a job in the outside sector, i.e. types for whom

$$
B\left(\frac{\bar{\varepsilon}+\theta e^{H}+a^{H}-\bar{y}}{\bar{\varepsilon}-\underline{\varepsilon}}\right)-\left(e^{H}\right)^{2}+(\tau+\Delta) e^{H}=\beta\left(\frac{\bar{\varepsilon}+\theta e^{O}-\bar{z}}{\bar{\varepsilon}-\underline{\varepsilon}}\right)-\left(e^{O}\right)^{2} .
$$

Substituting for $e^{H}$ and $e^{O}$ from Lemma 1, together with the parameters in our numerical example, and rearranging for $\tau$ yields

$$
\tau^{P}=\sqrt{25 \theta^{2}-4}-4 \theta-\frac{1}{2} .
$$

Fixing $\theta$, for any $\tau>\tau^{P}(\theta)$, the teacher's expected payoff from declining the government's offer and accepting a job in a high management private school is higher than her expected payoff from declining the government's offer and accepting a job in the outside sector.

Straightforward calculations show that $\tau^{P}-\tau^{G}$ is positive and increasing on [1.56, 5] which establishes that there exists a set $\mathcal{T}^{H} \equiv(\tau, \theta): \tau \geq \max \left\{\tau^{G}, \tau^{P}\right\}$. For any $(\tau, \theta)$ in this set, the expected payoff from declining the government offer and accepting a job in a high management private school exceeds both the payoff of accepting the government job and the expected payoff of declining and accepting a job in the outside sector.

Lemma 3. Assume that the government assigns the teacher to a low management public school. There exist functions

$$
\tau^{G^{\prime}}=\frac{36}{8 \theta+1}-2 \theta-\frac{1}{4}, \quad \tau^{O^{\prime}}=\sqrt{25 \theta^{2}-40}, \quad \text { and } \quad \tau^{P}=\sqrt{25 \theta^{2}-4}-4 \theta-\frac{1}{2}
$$

such that:

1. The teacher accepts the government's offer with probability $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{L}\right]$, where $\mathcal{T}^{L} \equiv$ $(\tau, \theta): \tau^{O^{\prime}}(\theta) \leq \tau \leq \tau^{G^{\prime}}(\theta)$.
2. The teacher declines the government's offer and accepts an offer from a private school with probability $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{H^{\prime}}\right]$, where $\mathcal{T}^{H^{\prime}} \equiv(\tau, \theta): \tau \geq \max \left\{\tau^{G^{\prime}}(\theta), \tau^{P}(\theta)\right\}$.

Proof. Analogous to Lemma 2.

Remark 1. Assume that the government assigns the teacher to an intermediate management public school. In the numerical example:

1. $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{H}\right]=0.741>\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{L}\right]=0.031$.
2. $\mathrm{E}\left[\tau+\Delta \mid(\tau, \theta) \in \mathcal{T}^{H}\right]=6.722>\mathrm{E}\left[\tau \mid(\tau, \theta) \in \mathcal{T}^{I}\right]=1.311$.
3. $\mathrm{E}\left[\left.\frac{\theta B}{2(\bar{\varepsilon}-\underline{\varepsilon})}+\frac{\tau+\Delta}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{H}\right]=8.851>\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{I}\right]=0.655$.

Proof. Calculated via numerical integration, using Lemmas 1 and 2. The Mathematica notebook file is available upon request.

Remark 2. Compare an intermediate management public school and a low management public school. In the numerical example

1. $\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{I}\right]=0.031>\operatorname{Pr}\left[(\tau, \theta) \in \mathcal{T}^{L}\right]=0.007$.
2. $\mathrm{E}\left[\tau \mid(\tau, \theta) \in \mathcal{T}^{I}\right]=1.311>\mathrm{E}\left[\tau \mid(\tau, \theta) \in \mathcal{T}^{L}\right]=0.545$.
3. $\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{I}\right]=0.655>\mathrm{E}\left[\left.\frac{\tau}{2} \right\rvert\,(\tau, \theta) \in \mathcal{T}^{L}\right]=0.301$.
4. $a^{I}=1>a^{L}=\frac{1}{2}$.

Proof. Calculated by numerical integration, using Lemmas 1 and 3. The Mathematica notebook file is available upon request.

ONLINE APPENDIX NOT INTENDED FOR PUBLICATION for Leaver, Lemos and Scur "Measuring and explaining management in schools: new approaches using public data," October 17, 2019

## B Additional tables and figures

Table B.1: Summary statistics

|  |  |  |  |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Mean | Standard <br> Deviation | 10 th <br> pct | 25 th <br> pct | 50 th <br> pct | 75 th <br> pct | 90 th <br> pct | N |
| School |  |  |  |  |  |  |  |  |
| Private school | 0.19 | $(0.39)$ | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 410200 |
| Rural | 0.32 | $(0.46)$ | 0.00 | 0.00 | 0.00 | 1.00 | 1.00 | 410209 |
| Student-teacher ratio | 17.47 | $(12.20)$ | 8.70 | 11.94 | 15.57 | 20.17 | 28.00 | 380244 |
| Enrolment total | 983.89 | $(789.31)$ | 222.00 | 439.00 | 813.00 | 1317.00 | 1861.00 | 394664 |
| Share of govt funding | 0.78 | $(0.32)$ | 0.14 | 0.68 | 0.95 | 1.00 | 1.00 | 377927 |
| Location: village | 0.13 | $(0.33)$ | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 410209 |
| Location: small town | 0.19 | $(0.39)$ | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 410209 |
| Location: town | 0.28 | $(0.45)$ | 0.00 | 0.00 | 0.00 | 1.00 | 1.00 | 410209 |
| Location: city | 0.26 | $(0.44)$ | 0.00 | 0.00 | 0.00 | 1.00 | 1.00 | 410209 |
| Location: large city | 0.15 | $(0.35)$ | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 410209 |
| Computers with internet | 0.87 | $(0.29)$ | 0.35 | 1.00 | 1.00 | 1.00 | 1.00 | 389971 |
| Student grade | 0.11 | $(0.32)$ | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 330163 |
|  |  |  |  |  |  |  |  |  |
| Students |  |  |  |  |  |  |  |  |
| Math score (PV1) | 457.47 | $(103.02)$ | 329.42 | 381.84 | 450.47 | 528.28 | 598.15 | 410701 |
| Reading score (PV1) | 465.67 | $(100.83)$ | 335.44 | 395.65 | 465.71 | 536.32 | 596.87 | 410701 |
| Science score (PV1) | 466.83 | $(100.85)$ | 339.18 | 394.01 | 463.20 | 537.52 | 602.14 | 410701 |
| Female student | 0.51 | $(0.50)$ | 0.00 | 0.00 | 1.00 | 1.00 | 1.00 | 410701 |
| Student age | 15.41 | $(0.53)$ | 14.33 | 15.25 | 15.58 | 15.83 | 15.92 | 410586 |
| Student: non-immigrant | 0.93 | $(0.26)$ | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 399606 |
| Student: second-gen | 0.04 | $(0.21)$ | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 399606 |
| Student: first-gen | 0.03 | $(0.16)$ | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 399606 |
| Socio-economic status index | 0.71 | $(0.48)$ | 0.12 | 0.31 | 0.65 | 1.02 | 1.36 | 175060 |

Table B.2: Above median people management and intermediate outcomes, public and private schools in Latin America

|  | $(1)$ <br> z-teacher <br> shortage | $(2)$ <br> z-teacher <br> motivation | $(3)$ <br> z-teacher <br> effort |
| :--- | :---: | :---: | :---: |
| Above median people | -0.186 | 0.403 | 0.060 |
|  | $(0.077)$ | $(0.073)$ | $(0.072)$ |
|  | $[0.015]$ | $[0.000]$ | $[0.406]$ |
| $R$-squared | 0.148 | 0.151 | 0.132 |
| Above 75th pct people | -0.245 | 0.482 | 0.065 |
|  | $(0.088)$ | $0.098)$ | $(0.086)$ |
|  | $[0.005]$ | $[0.000]$ | $[0.451]$ |
| $R$-squared | 0.150 | 0.152 | 0.132 |
| Observations | 3067 | 3074 | 3044 |
| School controls | Y | Y | Y |
| Country FE | Y | Y | Y |

Notes: The first row reports the coefficient from regressions of a binary indicator Above median people (coded to 1 if the school's PISA-based people management score is above the median, 0 otherwise) on the standardized index of three intermediate school outcomes: teacher shortage, teacher motivation and teacher effort. The second row reports coefficients from regressions of a binary indicator Above 75th pct people (coded to 1 if the school's PISA-based people management score is above 75 th percentile, 0 otherwise) on each of the intermediate school outcomes. The people management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include PISA school final weights and country fixed effects. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

Table B.3: Above median operations management and intermediate outcomes, public schools in Latin America

|  | $(1)$ <br> z-teacher <br> shortage | $(2)$ <br> z-teacher <br> motivation | $(3)$ <br> z-teacher <br> effort | $(4)$ <br> z-household <br> effort |
| :--- | :---: | :---: | :---: | :---: |
| Above median ops | -0.165 | 0.323 | 0.130 | 0.197 |
|  | $(0.082)$ | $(0.083)$ | $(0.075)$ | $(0.088)$ |
|  | $[0.044]$ | $[0.000]$ | $[0.083]$ | $[0.025]$ |
| $R$-squared | 0.0790 | 0.151 | 0.153 | 0.234 |
| Above 75pct ops | -0.121 | 0.498 | 0.221 | 0.348 |
|  | $(0.101)$ | $(0.103)$ | $(0.092)$ | $(0.121)$ |
| $R$-squared | $[0.233]$ | $[0.000]$ | $[0.016]$ | $[0.004]$ |
| Observations | 0.0759 | 0.164 | 0.156 | 0.240 |
| School controls | 2407 | 2414 | 2414 | 2414 |
| Country FE | Y | Y | Y | Y |

Notes: The row reports the coefficient from regressions of a binary indicator Above median ops (coded to 1 if the school's PISA-based people management score is above the median, 0 otherwise) on the standardized index of three intermediate school outcomes: teacher shortage, teacher motivation and teacher effort. The second row reports coefficients from regressions of a binary indicator Above 75th pct ops (coded to 1 if the school's PISA-based people management score is above 75 th percentile, 0 otherwise) on each of the intermediate school outcomes. The operations management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson [2008]. All specifications include PISA school final weights and country fixed effects. School controls include school location, student-teacher ratio, log of the number of students, share of government funding relative to total school funding, ratio of computers connected to the web as a proxy for school resources, and average student socio-economic status. For control variables, missing variables are replaced with a value of -99 and we include an indicator variable with a value of 1 for each imputed value, for each variable with imputed values dummies are added to the specifications.

Table B.4: Process of entering the public basic education teaching career in Latin American countries in 2012

| Country | Relevant Legislation in 2011-2012 | Eligibility to Apply | Process for Job offer and Allocation |
| :---: | :---: | :---: | :---: |
| Argentina <br> (Buenos <br> Aires) | Law 10.579, <br> Estatuto del Docente | Professional degree or equivalent in Teaching or Education in accordance to educational stage. <br> Candidates cannot be older than 50 years old. Foreign applicants must have at least 5 years of residency in the country. | 1) District government within Province announces vacancies. <br> 2) Candidate submits application along with supporting documentation. Candidate may choose a maximum of 3 districts per application (with no limit on the number of applications submitted). <br> 3) Decentralized classification tribunal (Tribunales de Clasificación Descentralizados) scores and ranks candidates based on supporting documentation (candidates applying to district of residency receive bonus points), following guidance and supervision of centralized classification tribunal at the provincial level. <br> 4) District government makes an offer to suitable candidates based on ranking, and offers permanent assignment after the candidate passes an assessment carried out during the first year of work. |
| Brazil <br> (Rio de Janeiro) | Municipal Law 2391/1995 <br> **Subsequent legislation was approved in 2013 (Plano de Cargos, Carreira e RemuneraçãoPCCR, Municipal Law № 5623/2013) | Professional degree in Teaching or Education. | 1) State government announces vacancies. <br> 2) Candidate registers in a specific district, pays an enrollment fee, takes state examination, and submits supporting documentation. <br> 3) State Office of Examination, Statistics and Public Service scores candidate's supporting documentation, conditional on passing state examination. <br> 4) State government publishes candidates' final score. <br> 5) Regional office under the state government (Coordenadoria Regional de Educação) makes an offer based on candidate's score, and offers permanent assignment after no more than 3 years of probationary period. |
| Chile | Law 19.070 passed in 1991. <br> Subsequent reform was carried in 2011 (Law 20.501) with focus on school principals. | Professional degree in Teaching or Education in accordance to educational stage. | 1) Municipal government announces vacancies. <br> 2) Candidate submits individual applications for each vacancy (with no limit on the number applications) along with supporting documentation. <br> 3) Municipal evaluation committee selects 2 to 5 candidates for each vacancy. <br> 4) Candidate presents a school work proposal to the committee. The proposal is ranked by the municipal evaluation committee (additional assessments may be requested). <br> 5) Municipal evaluation committee recommends candidates suitable for its vacancies, based on its ranking. <br> 6) Mayor makes an offer to suitable candidates, according to the recommendation received from the municipal evaluation committee. |
| Colombia | Law Decree 1278, 2002, Estatuto de Profesionalización Docente | Professional degree or equivalent. Non-educators are required to hold a specialization program on Pedagogy or similar field. | 1) Local government (Entidad Territorial) announces vacancies. <br> 2) Candidate registers, submits supporting documentation, and takes national examination. <br> 3) Local government committee evaluates the candidate based on a psycho-technical assessment, conditional on passing the national examination. <br> 4) Public Service National Authority decentralized committees (Comisión Nacional del Servicio Civil) score candidate's supporting documentation and carry out interviews. <br> 5) Public Service National Authority decentralized committees ranks candidates in each locality by educational stages and school modalities. Ranking is valid for 2 years. <br> 6) Local government makes an offer to suitable candidates based on ranking. <br> 7) Local government offers candidates permanent assignment after 1 year of probationary period, based on the school principal's evaluation. |

Table B.5: Process of entering the public basic education teaching career in Latin American countries in 2012

| Country | Relevant Legislation in 2011-2012 | Eligibility to Apply |  | Process for Job offer and Allocation |
| :---: | :---: | :---: | :---: | :---: |
| Costa Rica | 2005 Nueva <br> Carrera <br> Profesional Docente | Professional degree or equivalent in Teaching or Education. | 1) | National government announces vacancies. <br> Candidate submits application along with supporting documentation, and indicates region(s) of preference. <br> Public Service National Authority (Autoridad Nacional de Servicio Civil) ranks each candidate, allocates vacancies to suitable candidates, according to ranking and candidates' preferred region. National government offers candidate a permanent assignment after 3 months of probationary period, based on the school principal's evaluation. |
| Mexico | 2008, Alianza por la Calidad de la Educación. <br> **An education reform took place in 2013: Ley General del Servicio Profesional Docente | Professional degree or equivalent in Teaching or Education in accordance to educational stage. States have discretion to only accept candidates from specific teacher-training institutes or with a minimum time of residency in the State. | 1) 2) 3) 4) 5) | State government announces vacancies. <br> Candidate register, submits supporting documentation, and takes national examination. <br> National government grades examinations and sorts candidates into "Accepted", "Eligible" and "Not accepted" based on score. Cut-off points are determined by the Independent National Evaluation Body (Órgano de Evaluación Independiente con Carácter Federalista) and might differ across States. National government ranks "Accepted" and "Eligible" candidates based on their scores and publishes ranking. "Eligible" candidates will require additional training if accepted. State government makes an offer for a permanent assignment based on its ranking. |
| Peru | Law 29062, Ley de la Carrera Pública Magisterial. <br> **An education reform took place in Nov2012, and its regulation was later issued in Mar2013. | Professional degree in Teaching or Education in accordance to educational stage. Years of experience might be a requirement for candidates to vacancies in special or alternative education schools. | 4) | National government announces vacancies. <br> Candidate registers and takes national examination. <br> Candidate applies to a single school vacancy, conditional on meeting eligibility criteria and passing the national examination, and submits supporting documentation. <br> Evaluation committee (set up at school, municipality, local or regional level) scores candidate based on supporting documentation, school interview and classroom teaching practices. <br> Evaluation committee publishes a ranking of candidates per vacancy. <br> School makes an offer for a permanent assignment based on ranking provided by evaluation committee. |
| Uruguay | Estatuto Docente Regulation $\mathrm{N}^{\circ} 45$, approved by Act $\mathrm{N}^{\circ}$ 68, Resolution N ${ }^{\circ}$ 9, 1993; modified in 2008 | Professional degree in Teaching or Education in accordance to educational stage. | 1) | National council announces vacancies. <br> Candidate registers and submits supporting documentation. Candidate may apply to multiple vacancies in up to two municipalities but must submit individual applications. <br> Council sets up 3 Evaluation Committees per every 50 to 100 applicants. Committees scores candidates based on national level examination, classroom teaching practices, and supporting documentation. <br> Evaluation Committees rank candidates based on their total score. <br> Nacional Council makes an offer for a permanent assignment based on ranking for each vacancy. |

Figure B.1: Distribution of management scores, PISA 2012 vs WMS: operations


Note: Data for the World Management Survey index for all countries except for Mexico and Colombia can be found at www. worldmanagementsurvey.org. Distribution of operations management indices standardized within countries. Kernel density curves estimated using WMS sampling weights (calculated as the inverse probability of being interview on log of number of students, public status, and population density by state, province, or NUTS 2 region as a measure of location) for the WMS data and school final weights for the PISA data. Samples include both public and private secondary schools for both datasets, with the exception of Colombia where WMS data is only available to public primary schools. Number of WMS/PISA observations are as follow (WMS/PISA): Brazil $=510 / 561$, Canada $=129 / 770$, Colombia $=468 / 268$, Great Britain $=89 / 422$, Germany $=$ $102 / 158$, Italy $=284 / 926$, Mexico $=157 / 1327$, Sweden $=85 / 179$, United States $=263 / 136$.

Figure B.2: Distribution of management scores, PISA 2012 vs WMS: operations


Note: Data for the World Management Survey index for all countries except for Mexico and Colombia can be found at www.worldmanagementsurvey.org. Distribution of people management indices standardized within countries. Kernel density curves estimated using WMS sampling weights (calculated as the inverse probability of being interview on log of number of students, public status, and population density by state, province, or NUTS 2 region as a measure of location) for the WMS data and school final weights for the PISA data. Samples include both public and private secondary schools for both datasets, with the exception of Colombia where WMS data is only available to public primary schools. Number of WMS/PISA observations are as follow (WMS/PISA): Brazil $=510 / 561$, Canada $=129 / 770$, Colombia $=468 / 268$, Great Britain $=89 / 422$, Germany $=$ $102 / 158$, Italy $=284 / 926$, Mexico $=157 / 1327$, Sweden $=85 / 179$, United States $=263 / 136$.

Figure B.3: WMS score by quartile x country-specific student outcomes


Note: Reproduced from Bloom et al. [2015a]. Performance measures for 1002 observations: 472 for Brazil, 77 for Canada, 152 for India, 82 for Sweden, 86 for the UK and 133 for the US. At the time of writing, the authors of Bloom et al. [2015a] had conducted the WMS in 8 countries, the listed 6 plus Germany and Italy. The latter two countries are not included in this figure because data on student learning outcomes was not available to the authors.

Figure B．4：Cumulative distribution of people management：by country in Latin America


Notes：Cumulative distribution of the PISA－based people management index for private and public schools for each one of the 8 Latin American countries in the PISA 2012 dataset．The people management index is built out of the school questionnaire from PISA 2012 using the methodology from Anderson［2008］．Sample sizes are as follows：Argentina： 183 schools（ 63 private， 120 public）．Brazil： 561 schools（ 79 private， 482 public）．Chile： 201 schools（ 137 private， 64 public）．Colombia： 268 schools（ 62 private， 106 public）．Costa Rica： 158 schools（ 22 private 136 public）．Mexico： 1327 schools（196 private， 1131 public）．Peru： 207 schools（ 50 private， 157 public）．Uruguay： 164 schools（ 28 private， 136 public）．

Figure B.5: Standardized teacher shortage scores, by sector


Notes: Teacher Shortage index is built out of four PISA 2012 questions: "is your school's capacity to provide instruction hindered by any of the following issues? a lack of qualified [science, math, language, other subjects] teachers". The responses are scored as $1=$ not at all, $2=$ very little, $3=$ to some extent and $4=$ a lot. The index is built using the methodology in Anderson [2008]. Measures are standardized.

## C Data

## C. 1 Construction of the PISA-based indices

To construct a PISA-based school management index, we followed a three-step approach. First, we classified each of the PISA questions either under one of the WMS topics or under "not management". We were able to classify 532012 PISA questions into 14 WMS topics and using this mapping as a starting point, we further classified 322015 PISA questions into 12 WMS topics. For operations management, we classified 402012 PISA questions into 11 WMS topics and using this mapping as a starting point, we further classified 302015 PISA questions into 11 WMS topics. For people management, we classified 13 questions into 3 WMS topics using the 2012 questionnaire, and 2 questions into 1 WMS topic using the 2015 questionnaire. Table C. 6 provides a summary of the mapping for PISA 2012 and Prova Brasil 2013 used in this paper as well as a mapping for PISA 2015.

Table C.6: Mapping of Management Practices in Publicly Available Survey Data


Second, we manually assigned scores following the conceptual guidelines of the scoring grid of the World Management Survey, similar to the exercise conducted in the census-based management surveys such as the US Census Management and Organizational Practices Survey (MOPS), where values indicating best practices receive higher scores than values indicating poor practices. Values are normalized from 0 to $1 .{ }^{25}$

[^15]Third, to build the overall management index, and the operations and people management subindices, we follow Anderson [2008]. This methodology weights the impact of the included variables by the sum of their row in the inverse variance-covariance matrix, thereby assigning greater weight to questions that carry more "new information". Given that the importance (weight) of one questions is relative to the important of all others, we conservatily drop schools missing more than one management question (approximately $15 \%$ of schools are dropped, yet all countries are still included in the final sample). We also built the indices using alternative methods (straightforward standardization, factor analysis, including the Bartlett correction) which yielded similar results.

PISA 2015 has a reduced number of questions relative to the 2012 questions we used to measure people management. Several questions were moved to the new teacher questionnaire which was not mandatory for countries in 2015, preventing us from building an identically rich index across both years. For this reason, we focus on the richer 2012 data. A visual inspection of the distributions for the 2015 PISA-based management index for operations and people management shown in Figures C. 6 and C. 7 when compared to the distributions for 2012 indices in Figures B. 1 and B. 2 confirm that the 2012 indices, especially the people management index, are a better fit for this exercise.

We run two additional exercises to validate our index. First, we test whether the results are being driven by one specific question in the management index. To do this, we estimate the partial correlation of each of the 53 management questions on student performance, controlling for a partial index which takes into account all remaining management questions, standardized using Anderson [2008]. We find that the partial indices are positive and statistically significant throughout all individual regressions, suggesting that no single question is driving our results. Second, we test the importance of having both operations and people management questions in the index. This is to validate whether it is feasible to use the 2015 PISA data. To do this we run a regression where we include both operations and people management indices to the specification. This specification indicates whether each of the indices contain additively separable relevant information to explain student performance. We find that the coefficients remain positive and statistically significant across all subjects when including country fixed effects (same specification as Column (1) of Table 1), and remain positive and statistically significant in reading, marginally non-significant in math and science when fully specified (same specification as Column (3) of Table 1). Overall, these results suggest that both measures are still meaningful (tables are available upon request).

The list of questions included in the PISA 2012 management index and its mapping to the individual questions is described below.

Figure C.6: Distribution of management scores, PISA 2015 vs WMS: operations










$$
\text { PISA } 2015 \text { index } \mathbf{- = - = - \quad ~ W M S ~ i n d e x ~}
$$

Note: Data for the World Management Survey index for all countries except for Mexico and Colombia can be found at www.worldmanagementsurvey.org. Distribution of operations management indices standardized within countries. Kernel density curves estimated using WMS sampling weights (calculated as the inverse probability of being interview on log of number of students, public status, and population density by state, province, or NUTS 2 region as a measure of location) for the WMS data and school final weights for the PISA data. Samples include both public and private secondary schools for both datasets, with the exception of Colombia where WMS data is only available to public primary schools. Number of WMS/PISA observations are as follow $(\mathrm{WMS} / \mathrm{PISA}):$ Brazil $=510 / 421$, Canada $=129 / 562$, Colombia $=468 / 258$, Great Britain $=89 / 381$, Germany $=$ $102 / 156$, Italy $=284 / 291$, Mexico $=157 / 138$, Sweden $=85 / 192$, United States $=263 / 158$.

Figure C.7: Distribution of management scores, PISA 2015 vs WMS: people


CAN


GER


SWE


COL


ITA


USA

——PISA 2015 index --=-. WMS index

Note: Data for the World Management Survey index for all countries except for Mexico and Colombia can be found at www.worldmanagementsurvey.org. Distribution of people management indices standardized within countries. Kernel density curves estimated using WMS sampling weights (calculated as the inverse probability of being interview on log of number of students, public status, and population density by state, province, or NUTS 2 region as a measure of location) for the WMS data and school final weights for the PISA data. Samples include both public and private secondary schools for both datasets, with the exception of Colombia where WMS data is only available to public primary schools. Number of WMS/PISA observations are as follow $(\mathrm{WMS} / \mathrm{PISA}):$ Brazil $=510 / 421$, Canada $=129 / 562$, Colombia $=468 / 258$, Great Britain $=89 / 381$, Germany $=$ $102 / 156$, Italy $=284 / 291$, Mexico $=157 / 138$, Sweden $=85 / 192$, United States $=263 / 158$.

|  |  | PISA 2012 |  |  |
| :--- | :--- | :--- | :--- | :--- |


| WMS questions | Questions | Var. name in questionnari e | Value label | V | MGMT <br> score |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 4) Adopting Educational Best Practices |  |  |  |  |  |
| a) How does the school encourage incorporating new teaching practices into the classroom? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I promote teaching practices based on recent educational research. | SC34Q05 | $\begin{aligned} & \text { Did not occur } \\ & \hline 1-2 \text { times during } \\ & \text { the year } \\ & \hline \end{aligned}$ | 2 | 0.00 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
| b) How are these learning or new teaching practices shared across teachers? What about across grades or subjects? How does sharing happen across schools (community, statewide etc), if at all? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I set aside time at faculty meetings for teachers to share ideas or information from in-service activities. | SC34Q18 | $\begin{aligned} & \frac{\text { Did not occur }}{1-2 \text { times during }} \\ & \text { the year } \\ & \hline \end{aligned}$ | 1 | 0.00 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
|  | Which of the following measures aimed at quality assurance and improvement do you have in your school? Answer: Teacher mentoring. | SC39Q08 | Yes | , | 1.00 |
|  |  |  | No | 2 | 0.00 |
|  | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I lead or attend in-service activities concerned with instruction. | SC34Q17 | Did not occur | 1 | 0.00 |
|  |  |  | 1-2 times during the year | 2 | 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
| 5) Continuous Improvement |  |  |  |  |  |
| a) When problems (e.g. within school/ teaching tactics/ etc.) do occur, how do they typically get exposed and fixed? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: When a teacher has problems in his/her classroom, I take the initiative to discuss matters. | SC34Q07 | Did not occur | 1 | 0.00 |
|  |  |  | 1-2 times during the year | 2 | 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
| c) Who within the school gets involved in changing or improving process? How do the different staff groups get involved in this? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I engage teachers to help build a school culture of continuous improvement. | SC34Q11 | Did not occur | 1 | 0.00 |
|  |  |  | 1-2 times during the year | 2 | 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |




| WMS questions | Questions | Var. name in questionnari e | Value label | V a l u | MGMT score |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 14) Clarity and Comparability of Targets |  |  |  |  |  |
| a) If I asked one of your staff members directly about individual targets, what would they tell me? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I ensure that teachers work according to the school's educational goals. | SC34Q04 | $\begin{aligned} & \frac{\text { Did not occur }}{1-2 \text { times during }} \\ & \text { the year } \\ & \hline \end{aligned}$ | 2 | 0.00 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
|  | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I refer to the school's academic goals when making curricular decisions with teachers. | SC34Q15 | Did not occur | 1 | 0.00 |
|  |  |  | 1-2 times during the year | 2 | 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
|  | Which of the following measures aimed at quality assurance and improvement do you have in your school? Answer: Written specification of the school's curricular profile and educational goals. | SC39Q01 | Yes | 1 | 1.00 |
|  |  |  | No | 2 | 0.00 |
|  | Which of the following measures aimed at quality assurance and improvement do you have in your school? Answer: Written specification of student performance standards. | SC39Q02 | Yes | 1 | 1.00 |
|  |  |  | No | 2 | 0.00 |
| c) How do people know about their own performance compared to other people's performance? | In your school, are achievement data used in any of the following <accountability procedures>? Answer: Achievement data are posted publicly (e.g. in the medil). | SC19Q01 | Yes | 1 | 1.00 |
|  |  |  | No | 2 | 0.00 |
|  | In your school, are achievement data used in any of the following <accountability procedures>? Answer: Achievement data are tracked over time by an administrative authority. | SC19Q02 | Yes | 1 | 1.00 |
|  |  |  | No | 2 | 0.00 |
|  | 15) Rewarding High Performers |  |  |  |  |
| a) How does your evaluation system work? What proportion of your employees' pay is related to the results of this review? | Please indicate the frequency of the following activities and behaviours in your school during <the last academic year>. Answer: I evaluate the performance of staff. | SC34Q22 | Did not occur | 1 | 0.00 |
|  |  |  | 1-2 times during the year | 2 | 0.20 |
|  |  |  | 3-4 times during the year | 3 | 0.40 |
|  |  |  | Once a month | 4 | 0.60 |
|  |  |  | Once a week | 5 | 0.80 |
|  |  |  | More than once a week | 6 | 1.00 |
|  | Are assessments of students in <national modal grade for 15-year-olds> used to make judgements about teachers' effectiveness? | SC18Q06 | Yes | 1 | 1.00 |
|  |  |  | No | 2 | 0.00 |
| b) Are there any nonfinancial or financial bonuses/ rewards for the best performers across all staff groups? How does the bonus system work (for staff and teachers)? | -To what extent have appraisals of and/or feedback to teachers directly led a change in salary? | SC31Q01 | No change | 1 | 0.00 |
|  |  |  | Small change | 2 | 0.33 |
|  |  |  | Moderate change | 3 | 0.66 |
|  |  |  | Large change | 4 | 1.00 |
|  | To what extent have appraisals of and/or feedback to teachers directly led a financial bonus or another kind of monetary reward? | SC31Q02 | No change | 1 | 0.00 |
|  |  |  | Small change | 2 | 0.33 |
|  |  |  | Moderate change | 3 | 0.66 |
|  |  |  | Large change | 4 | 1.00 |
|  | To what extent have appraisals of and/or feedback to teachers directly led a public recognition from you? | SC31Q05 | No change | 1 | 0.00 |
|  |  |  | Small change | 2 | 0.33 |
|  |  |  | Moderate change | 3 | 0.66 |
|  |  |  | Large change | 4 | 1.00 |


C. 2 Construction of teacher shortage, teacher motivation, teacher effort, and household effort indices

We use the Anderson [2008] methodology to build each intermediate teacher outcomes index below.

| PISA 2012 |  |  |  |
| :---: | :---: | :---: | :---: |
| Questions | Var. name in questionnarie | Value label | Value |
| Teacher Shortage |  |  |  |
| Is your school's capacity to provide instruction hindered by any of the following issues? Answer: A lack of qualified science teachers. | SC14Q01 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Is your school's capacity to provide instruction hindered by any of the following issues? Answer: A lack of qualified mathematics teachers. | SC14Q02 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Is your school's capacity to provide instruction hindered by any of the following issues? Answer: A lack of qualified <test language> teachers. | SC14Q03 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Is your school's capacity to provide instruction hindered by any of the following issues? Answer: A lack of qualified teachers of other subjects. | SC14Q04 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Teacher Motivation |  |  |  |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teachers' low expectations of students. | SC22Q13 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teachers not meeting individual students' needs. | SC22Q14 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Discussed their child's behaviour on the initiative of one of their child's teachers. | SC25Q02 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Discussed their child's progress on the initiative of one of their child's teachers. | SC25Q04 | Percentage |  |
| Think about the teachers in your school. How much do you agree with the following statements? Answer: The morale of teachers in this school is high. | SC26Q01 | Strongly agree |  |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| Think about the teachers in your school. How much do you agree with the following statements? Answer: Teachers work with enthusiasm. | SC26Q02 | Strongly agree |  |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| Think about the teachers in your school. How much do you agree with the following statements? Answer: Teachers take pride in this school. | SC26Q03 | Strongly agree | 2 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| Think about the teachers in your school. How much do you agree with the following statements? Answer: Teachers value academic achievement. | SC26Q04 | Strongly agree | 1 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| How much do you agree with these statements about teachers in your school? Answer: Mathematics teachers are interested in trying new methods and teaching practices. | SC27Q01 | Strongly agree | 1 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| How much do you agree with these statements about teachers in your school? <br> Answer: There is a preference among mathematics teachers to stay with well-known methods and practices. | SC27Q02 | Strongly agree |  |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| How much do you agree with these statements about teachers in your school? | SC28Q01 | Strongly agree | 1 |
| Answer: There is consensus among mathematics teachers that academic achievement must be kept as high as possible. |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |

App. 22

| PISA 2012 |  |  |  |
| :---: | :---: | :---: | :---: |
| Questions | Var. name in questionnarie | Value label | Value |
| Teacher Motivation |  |  |  |
| How much do you agree with these statements about teachers in your school? Answer: There is consensus among mathematics teachers that it is best to adapt academic standards to the students' levels and needs. | SC28Q02 | Strongly agree | 1 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| How much do you agree with these statements about teachers in your school? Answer: There is consensus among mathematics teachers that the social and emotional development of the students is as important as their acquisition of mathematical skills and knowledge in mathematics classes. | SC29Q01 | Strongly agree | 1 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| How much do you agree with these statements about teachers in your school? Answer: There is consensus among mathematics teachers that the development of mathematical skills and knowledge in students is the most important objective in mathematics classes. | SC29Q02 | Strongly agree | 1 |
|  |  | Agree | 2 |
|  |  | Disagree | 3 |
|  |  | Strongly disagree | 4 |
| Teacher Effort |  |  |  |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teachers having to teach students of heterogeneous ability levels within the same class. | SC22Q11 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teacher absenteeism. | SC22Q15 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teachers being too strict with students. | SC22Q17 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Teachers being late for classes. | SC22Q18 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Household Effort |  |  |  |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Student truancy. | SC22Q01 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Students skipping classes. | SC22Q02 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Students arriving late for school. | SC22Q03 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Students not attending compulsory school events (e.g. sports day) or excursions. | SC22Q04 | Not at all | I |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Students lacking respect for teachers. | SC22Q05 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Disruption of classes by students. | SC22Q06 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Students intimidating or bullying other students. | SC22Q08 | Not at all |  |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |

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| PISA 2012 |  |  |  |
| :---: | :---: | :---: | :---: |
| Questions | Var. name in questionnarie | Value label | Value |
| Household Effort |  |  |  |
| In your school, to what extent is the learning of students hindered by the following phenomena? Answer: Poor student-teacher relations. | SC22Q10 | Not at all | 1 |
|  |  | Very little | 2 |
|  |  | To some extent | 3 |
|  |  | A lot | 4 |
| Which statement below best characterises parental expectations towards your school? | SC24Q01 | There is constant pressure from many parents, who expect our school to set very high academic standards and to have our students achieve them. | 1 |
|  |  | Pressure on the school to achieve higher academic standards among students comes from a minority of parents. | 2 |
|  |  | Pressure from parents on the school to achieve higher academic standards among students is largely absent. | 3 |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Discussed their child's behaviour with a teacher on their own initiative. | SC25Q01 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Discussed their child's progress with a teacher on their own initiative. | SC25Q03 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Volunteered in physical activities, e.g. building maintenance, carpentry, gardening or yard work. | SC25Q05 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Volunteered in extra-curricular activities, e.g. book club, school play, sports, field trip. | SC25Q06 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Volunteered in the school library or media centre. | SC25Q07 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Assisted a teacher in the school. | SC25Q08 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Appeared as a guest speaker. | SC25Q09 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Participated in local school <government>, e.g. parent council or school management committee. | SC25Q10 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Assisted in fundraising for the school. | SC25Q11 | Percentage |  |
| During <the last academic year>, what proportion of students' parents participated in the following school-related activities? Answer: Volunteered in the school <canteen>. | SC25Q12 | Percentage |  |

## C. 3 Construction of the Prova Brasil-based school management index

To construct the Prova Brasil-based school management index, we followed the three steps as detailed in the construction of the PISA-based index. However, as we map variables from both the school director and teacher questionnaires, we take one further step: we collapse the teacher dataset at the school level, taking the average of all teacher responses by school, combine the school principal responses, and compute the school level index. The WMS-Prova Brasil 2013 mapping is detailed below. For a harmonized version of the Prova Brasil mapping across 2007 to 2017, see Adelman et al. [2019].


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| WMS questions | Questions |  |  |  |
| :--- | :--- | :--- | :--- | :--- | within the

classroom?

| c) What about students, how does the school ensure they are engaged in their own learning? How are parents incorporated in this process? | Indique com qual frequência são desenvolvidas as seguintes | Principal: | A | Nunca. | 0.00 |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | atividades para minimizar as faltas dos alunos neste ano e nesta | TX_RESP_Q045 | B | Algumas vezes. | 0.33 |
|  | escola: Os professores conversam com os alunos para tentar |  | C | Frequentemente. | 0.66 |
|  | solucionar o problema. |  | D | Sempre ou quase sempre. | 1.00 |
|  | Indique com qual frequência são desenvolvidas as seguintes atividades para minimizar as faltas dos alunos neste ano e nesta escola: Os pais/responsáveis são avisados por comunicação da escola. | Principal: | A | Nunca. | 0.00 |
|  |  | TX_RESP_Q046 | B | Algumas vezes. | 0.33 |
|  |  |  | C | Frequentemente. | 0.66 |
|  |  |  | D | Sempre ou quase sempre. | 1.00 |
|  | Indique com qual frequência são desenvolvidas as seguintes atividades para minimizar as faltas dos alunos neste ano e nesta escola: Os pais/responsáveis são chamados à escola para conversar sobre o assunto em reunião de pais. | Principal: | A | Nunca. | 0.00 |
|  |  | TX_RESP_Q047 | B | Algumas vezes. | 0.33 |
|  |  |  | C | Frequentemente. | 0.66 |
|  |  |  | D | Sempre ou quase sempre. | 1.00 |
|  | Indique com qual frequência são desenvolvidas as seguintes atividades para minimizar as faltas dos alunos neste ano e nesta escola: Os pais/responsáveis são chamados à escola para conversar sobre o assunto individualmente. | Principal: | A | Nunca. | 0.00 |
|  |  | TX_RESP_Q048 | B | Algumas vezes. | 0.33 |
|  |  |  | C | Frequentemente. | 0.66 |
|  |  |  | D | Sempre ou quase sempre. | 1.00 |
|  | Indique com qual frequência são desenvolvidas as seguintes atividades para minimizar as faltas dos alunos neste ano e nesta escola: A escola envia alguém à casa do aluno. | Principal: | A | Nunca. | 0.00 |
|  |  | TX_RESP_Q049 | B | Algumas vezes. | 0.33 |
|  |  |  | C | Frequentemente. | 0.66 |
|  |  |  | D | Sempre ou quase sempre. | 1.00 |
|  | Na sua percepção, os possíveis problemas de aprendizagem dos alunos das série(s) ou ano(s) avaliado(s) ocorrem, nesta escola, devido à/ao(s): Falta de assistência e acompanhamento dos pais na vida escolar do aluno. | Teacher: <br> TX RESP Q78 | A | Sim | 0.00 |
|  |  |  | B | Não | 1.00 |
|  | Na sua percepção, os possíveis problemas de aprendizagem dos alunos das série(s) ou ano(s) avaliado(s) ocorrem, nesta escola, devido à/ao(s): Desinteresse e falta de esforço do aluno. | Teacher: | A | Sim | 0.00 |
|  |  | TX_RESP_Q80 | B | Não | 1.00 |
|  | Na sua percepção, os possíveis problemas de aprendizagem dos alunos das série(s) ou ano(s) avaliado(s) ocorrem, nesta escola, devido à/ao(s): Alto índice de faltas por parte dos alunos. | Teacher: | A | Sim | 0.00 |
|  |  | TX_RESP_Q82 | B | Não | 1.00 |
|  | 4) Adopting Educational Best Practices |  |  | Operations Managemen |  |
| a) How does the school encourage incorporating new teaching practices into the classroom? | Qual foi a quantidade de docentes desta escola que participou das atividades de formação continuada que você organizou nos últimos dois anos? | Principal: TX_RESP_Q027 | A | Não foram organizadas atividades de formação | 0.00 |
|  |  |  | B | Poucos professores. | 0.25 |
|  |  |  | C | Um pouco menos da metade dos professores. | 0.50 |
|  |  |  | D | Um pouco mais da metade dos professores. | 0.75 |
|  |  |  | E | Quase todos ou todos os professores. | 1.00 |
|  | Nesta escola e neste ano, indique a frequência com que: $\mathrm{O}(\mathrm{A})$ diretor(a) estimula atividades inovadoras. | Teacher: <br> TX_RESP_Q65 | A | Nunca. | 0.00 |
|  |  |  | B | Algumas vezes. | 0.33 |
|  |  |  | C | Frequentemente. | 0.66 |
|  |  |  | D | Sempre ou quase sempre. | 1.00 |





[^0]:    ${ }^{1}$ For example, see Wössmann [2016] for a review of education systems research using large, cross-country surveys.

[^1]:    ${ }^{2}$ This assumption echoes the observation made by Baker et al. [1988] that compensation plans featuring explicit financial rewards seldom account for all of a worker's rewards.

[^2]:    ${ }^{3}$ See also Dohmen and Falk [2010] who briefly sketch out theoretical reasons why fixed wage contracts and piece rates might be expected to have different impacts on sorting by ability.

[^3]:    ${ }^{4}$ Our approach follows the spirit of the re-casting of the original phone-based World Management Survey into the US Census Management and Organizational Practices Survey (MOPS) administered to the population of US manufacturing establishments as a self-reported questionnaire [Bloom et al., 2019a]. The MOPS has been replicated in a number of other countries. Its questions follow the WMS topics and look to measure similar practices, but with self-reported answers.
    ${ }^{5}$ See Bloom and Van Reenen [2007] for the survey's inception and Bloom et al. [2016] for a recent review.

[^4]:    ${ }^{6}$ We replicate the primary figure from Bloom et al. [2015a] in Figure B. 3 in Appendix B. It plots the school-level student learning outcomes by each quartile of school management score.
    ${ }^{7}$ We provide details to enable replication in Appendix C
    ${ }^{8}$ Our main focus is on the 2012 data because that survey wave contains a richer set of questions, particularly relating to people management. See Appendix C for a mapping of the 2015 PISA data.
    ${ }^{9}$ The set of WMS topics that had matching PISA questions is detailed in Appendix C.

[^5]:    ${ }^{10}$ We also built the indices using alternative methods (straightforward standardization, factor analysis, and factor analysis with Bartlett correction) which yielded similar results.
    ${ }^{11}$ In the 2012 PISA the dataset included a PISA-built 'leadership and management' measure. This is distinctively different from ours, as it was based off a section of the questionnaire that was titled 'management' and contained only a small subset of questions. This index fails to take advantage of the full questionnaire and the information available elsewhere that also speaks to managerial practices used in the schools. More pertinently, PISA's measure does not compare well to the (empirically robust) management index derived from the World Management Survey (see Liberto et al. [2015]).

[^6]:    ${ }^{12}$ Independent researchers conducted the WMS in Colombia and Mexico during 2015, with guidance and supervision from the original WMS team. These data were not available to Bloom et al. [2015a]. India is not included in Figure 1 because it did not participate in PISA in 2012.
    ${ }^{13}$ We show comparisons between the WMS and PISA sub-indices for operations management in Figure B. 1 and people management in in Figure B. 2 in Appendix B and confirm that the distributions are consistent for both sub-indices.
    ${ }^{14}$ Prova Brasil was also administered in 2013.

[^7]:    ${ }^{15}$ See Jerrim et al. [2017] for a thorough review of how to best use PISA scores and survey weights.
    ${ }^{16}$ PISA is standardized across years and countries such that the mean is 500 and the standard deviation is 100 .

[^8]:    ${ }^{17}$ In principle, parents could play a further role by selecting between schools. Since our PISA data cannot speak to this issue, we leave the analysis of management-induced household selection for future work. Note that we assume management practices change effective labour inputs. In their study of the IT industry, Schivardi and Schmitz [2019] assume that management is an additional input, alongside capital and labour, in an approach that they term "management as a production technology".

[^9]:    ${ }^{18}$ Note that the basic production function is the same across schools; management practices matter by affecting which $\theta$-types are hired, the teacher's choice of $e$ (which depends on which $\tau$-types are hired), and the household's choice of $a$.

[^10]:    ${ }^{19}$ We show empirical CDFs by country in Latin America in Figure B. 4 in Appendix B. The private sector CDF first order stochastically dominates the public sector CDF in Colombia, Peru, Costa Rica and Mexico.
    ${ }^{20}$ In assuming this timing, we abstract from applicant choice between schools in the public sector. As Table B. 4 and B. 5 in Appendix B, the degree to which teachers can choose among public schools varies across Latin America, yet our model generally fits the reality in these countries.

[^11]:    ${ }^{21}$ These predictions are based on the numerical example illustrated in Figure 7 and are derived (via numerical integration) in Remarks 1 and 2 in Appendix A. We present the main results with our preferred specification, but include additional variations of the main explanatory variable in Table B. 3 in Appendix B. The results are robust to alternative specifications.
    ${ }^{22}$ We describe how this teacher shortage index, and the other intermediate outcome indices for teacher motivation, teacher effort, and household involvement, are constructed in Appendix C. All indices are standardized.

[^12]:    ${ }^{23}$ Consistent with this, by far the most common answer given by principals in both sectors is 'not at all', as reflected in the density of the standardized score in Figure B. 5 in Appendix B.

[^13]:    ${ }^{24}$ We present the main results with our preferred specification, but include additional variations of the main explanatory variable in Tables B. 2 in Appendix B. The results are robust to alternative specifications.

[^14]:    Note: Number of observations: 15,196 schools from 65 countries available in PISA 2012 data. Student outcomes are estimated using five plausible values and collapsed at the school level using PISA's senate weights. Quartiles of management are built at the country level. Test scores are presented as deviations from the global mean.

[^15]:    ${ }^{25}$ MOPS has since been replicated in a number of other countries. Its questions follow the WMS topics and look to measure similar practices, but with self-reported answers.

