

**THE GOVERNANCE OF NON-PROFITS AND THEIR SOCIAL IMPACT:
EVIDENCE FROM A RANDOMIZED PROGRAM IN HEALTHCARE IN THE DEMOCRATIC
REPUBLIC OF CONGO**

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ABSTRACT

How can non-profit organizations improve their governance to increase their social impact? This study examines the effectiveness of a bundle of governance mechanisms—consisting of social performance-based incentives combined with auditing and feedback—in the context of a randomized governance program conducted in the Democratic Republic of Congo’s healthcare sector. Within the program, a set of health centers were randomly assigned to a governance treatment while others were not. We find that the governance treatment leads to i) higher operating efficiency and ii) improvements in social performance (measured by a reduction in the occurrence of stillbirths and neonatal deaths). Furthermore, we find that funding is not a substitute for governance—health centers that only receive funding increase their scale, but do not show improvements in operating efficiency nor social performance. Overall, our results suggest that governance plays an important role in achieving the non-profits’ objectives and increasing the social impact of the funds invested.

Keywords: non-profit governance; non-profit organizations; social impact; healthcare; Sustainable Development Goals (SDGs); developing countries; randomized experiment.

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1. Introduction

Every year, considerable efforts and large amounts of funds are invested in social and environmental causes aiming to achieve the United Nations' Sustainable Development Goals (SDGs). Those include ending poverty, reducing hunger, promoting healthy lives and well-being, reducing inequalities, promoting the development of sustainable cities and communities, addressing climate change, and protecting the world's biodiversity, among others. Accounting for about one-third of total employment in the social sector, non-profit organizations represent a large part of the global economy and constitute a major player in achieving the SDGs (United Nations 2018a).¹

To help tackle these grand societal challenges and achieve the SDGs, it is important to understand how to improve the effectiveness of non-profit organizations. Arguably, this is a critical question not only for academics but also, and more importantly, for the non-profit organizations themselves and the beneficiaries of their services—affected individuals, communities, and the natural environment. What is more, the efficient use of funds is also important for donors and impact investors who aim to make a difference in this world and maximize their funds' social impact. In sum, understanding what mechanisms are available and effective in improving the social impact of non-profit organizations is important for the organizations per se, the donors and impact investors, as well as society at large and the natural environment.

The question of how to improve the governance of non-profits is difficult to answer, both theoretically and empirically. The insights from the existing literature on governance offer only limited guidance in this regard since it has largely focused its attention on for-profit organizations, which are fundamentally different from non-profit organizations.²

¹ For example, in the U.S., the non-profit sector represents 10.2% of the private sector's labor force in 2017 (Salamon and Newhouse 2020).

² By their very nature, non-profits pursue social as opposed to financial objectives. In this regard, an inherent feature of non-profits is the “non-distribution constraint” (Hansmann 1980)—that is, non-profits are not allowed to distribute profits to donors or employees; instead, any surplus they generate must be retained and devoted to their social objectives. A direct implication of this non-distribution constraint is that non-profits do not have owners. The investors who fund non-profits, through donations, do not have any claim to the non-profits' revenues and assets, nor do they

From a theoretical perspective, it is unclear what governance mechanisms are available and effective for non-profits to maximize the social impact of the funds invested, as well as when and where they matter more or less. In order to be effective, appropriate governance mechanisms need to take into account the unique nature of and obstacles faced by non-profit organizations. First, non-profits are often cash-constrained, their employees might be purpose-driven and intrinsically motivated, and key (non-financial) performance metrics might be difficult to measure and evaluate. Second, many non-profits operate in countries that are resource-constrained (e.g., in terms of financial resources and qualified personnel). In this regard, the provision of additional funding is likely helpful. However, funding per se need not be sufficient to bring about tangible improvements in the non-profits' social impact. In particular, their managers and employees may lack the necessary knowledge on how to improve the organization's operating efficiency and service quality. Accordingly, even if they had access to more funding, they might not be able to put this funding to good use without better governance and management practices. For example, in the context of the health sector in the Democratic Republic of Congo (DRC), the WHO has identified several sources of inefficiencies such as the lack of strategic and managerial planning, inadequate priorities in resource allocation, lack of transparency, lack of managerial competencies, and insufficient medical training of health workers (WHO 2015). These sources of inefficiencies are unlikely unique to the health sector in the DRC, but rather a common challenge found across sectors and across (low-income) countries.

What is more, these countries may lag behind several of the SDGs. Indeed, taking the example of the healthcare sector in the DRC, the promotion of healthy lives and well-being is a complex issue that not only requires access to effective healthcare services, but also access to affordable and clean energy, clean water and sanitation, education, and the achievement of several other SDGs. Achieving the SDGs is a complex task that likely takes a concerted effort by non-profits, for-profits, governments, and society more generally. As such, it is unclear whether and to what extent the provision of funding and better governance

have any control rights over the organization. Accordingly, many of the governance tools available to for-profits (e.g., managerial ownership) are not available to non-profits.

can truly move the needle and bring about substantial improvements in operating efficiency and service quality.

From an empirical perspective, making ground on these questions is challenging. There are two main obstacles. First, it is difficult to obtain fine-grained microdata on non-profit organizations, their governance, as well as their social impact. Second, even if detailed microdata were available, the adoption of governance practices is likely endogenous with respect to organizational outcomes—that is, unobservables may drive a spurious relationship between the adoption of different governance mechanisms and organizational outcomes. Addressing the endogeneity of governance requires a source of exogenous variation in the adoption of governance practices.

To overcome these obstacles, we study the governance of non-profits in the context of a randomized governance program implemented in the healthcare sector of the Democratic Republic of Congo (DRC). In the DRC, primary healthcare services—and especially maternity and childbirth services—are administered in non-profit health centers spread across the country. The randomized governance program we exploit in this paper was administered in about 1,000 health centers at the beginning of 2017. Health centers in the program were randomly assigned to a treatment group and control group, respectively. While health centers in both groups received funding from the program, only those in the treatment group were subject to a “governance treatment” consisting of social performance-based incentives combined with auditing and feedback (A&F).³ Hence, by design, this randomized governance program provides an ideal setup to study how the adoption of governance practices affects health centers’ outcomes (e.g., their operating efficiency and social impact), holding everything else, including funding, constant.

Using a difference-in-differences methodology, we find that, within a 10-quarter period following the treatment, health centers in the treatment group experience significant improvements in both operating efficiency (captured by an increase in the number of services provided per employee) and social

³ Social performance-based incentives are a form of pay-for-social-performance. Specifically, additional funding is provided to the non-profit organization contingent on the organization’s social performance. See Section 3.2 for details.

performance (captured by a reduction in the probability of stillbirths and neonatal deaths, respectively). These findings suggest that the adoption of governance mechanisms (in the form of social performance-based incentives and A&F) is beneficial to non-profits and contributes to their ability to achieve their intended social impact.

We further document that the treatment effect takes several quarters to materialize, consistent with a “learning” interpretation—that is, the health center staff progressively learn how to improve their operations from the quarterly rounds of feedback, while having incentives to act on this feedback. When we examine the heterogeneity in the treatment effect, we find further evidence that points toward the importance of learning. Specifically, we find that the treatment effect is stronger in areas with a lower density of health centers (that is, areas where it is harder to learn and adopt best practices from their peers). Moreover, we find that i) the increase in operating efficiency is higher for health centers with lower ex ante operating efficiency, and ii) the decrease in infant mortality is more pronounced for health centers with higher ex ante mortality. Collectively, these findings indicate that the improvements brought about by the treatment are larger when there is more to improve and learn to begin with.

Finally, we examine whether funding can serve as a substitute for governance. This question is not only relevant from a societal perspective but also from an implementation and policy perspective. Indeed, transferring money is relatively straightforward, while it is more challenging to implement better governance and management practices. To shed light on this question, we compare health centers in- and outside the governance program. We refer to the latter group as the “outside group.” Unlike the treatment group (i.e., health centers that receive both funding and the governance treatment) and the control group (i.e., health centers that only receive funding), health centers in the outside group receive neither. Accordingly, by comparing the treatment group versus the outside group, we can estimate the impact of the combination of ‘funding and governance’ on health center outcomes. Similarly, by comparing the control group versus the outside group, we can measure the impact of ‘funding’ as standalone. This analysis reveals that funding is not a substitute for governance—health centers that only receive funding increase their scale, but do not show improvements in operating efficiency nor social performance. In contrast, health centers

that receive both funding and the governance treatment improve both their scale as well as their operating efficiency and social performance.

Overall, our results suggest that governance plays an important role in contributing to non-profits' objectives and increasing their social impact, and that funding alone does not serve as a substitute for governance. Naturally, we caution that our findings are specific to the healthcare sector in a low-income country, and hence need not generalize to the overall non-profit sector. Nevertheless, studying this specific context is important—it speaks to the effective pursuit of one of the United Nations' SDGs, namely the promotion of healthy lives and well-being, especially with respect to children's health. In the past years, major progress has been made in improving the health of millions of people, reducing premature deaths and increasing their life expectancy. In particular, according to the World Health Organization (WHO), the mortality rate of children under the age of five decreased from 9.3% (i.e., 93 deaths per 1,000 live births) in 1990, to 3.9% in 2017 (WHO 2018). This corresponds to a worldwide decline from 12.6 million deaths in 1990, to 5.4 million in 2017. Yet, despite this overall progress, stark disparities exist across regions and countries. In particular, Sub-Saharan Africa remains the region with the highest under-5 mortality rate in the world, with one child in thirteen dying before their fifth birthday—this is fourteen times higher than in high-income countries (WHO 2018). A similar picture arises with the mortality rate of infants under one-year old. As Figure A1 shows, the DRC is among the countries with the highest infant mortality rates in the world. An estimated 7% of infants died within the first year of their life in 2017, compared to 0.4% in France and 0.6% in the U.S. (United Nations 2018b). Reducing children mortality is of foremost importance, and this study helps inform how governance can contribute to this objective.

2. The role of governance in improving non-profits' operating efficiency and their social impact

In the following, we discuss the potential effectiveness of a bundle of governance mechanisms—consisting of social performance-based incentives combined with auditing and feedback (A&F)—for improving the non-profit organizations' operating efficiency and social outcomes. This governance bundle is used in the randomized governance program we study in the empirical analysis.

Social performance-based incentives

To motivate managers and employees to use the non-profits' resources efficiently and maximize their social impact, donors (and other impact investors) can provide social performance-based incentives. That is, additional funding is provided to the non-profit organization conditional on meeting specific social criteria. As such, social performance-based incentives differ from the more traditional financial incentives in two ways: i) the additional funding is tied to social criteria in lieu of financial criteria, and ii) the direct beneficiary of social performance-based incentives is the non-profit organization itself as opposed to the managers and employees, respectively.

Social performance-based incentives may motivate non-profit managers and employees in two ways. First, by providing such incentives, non-profits can leverage managers' and employees' intrinsic motivation to obtain additional funding for the non-profit's cause.⁴ Second, social performance-based incentives may provide direct benefits to the managers and employees. Indeed, the additional funding allows non-profit organizations to, e.g., upgrade their equipment, provide training to their employees, pay higher wages, extend the scope of their services, all of which increase the attractiveness of the workplace and, as a result, can have a motivational effect on individuals' work behavior.⁵ On top of the motivational aspect, providing employees and managers with training, upgraded equipment, etc., may empower them to further increase their productivity.

While the provision of social performance-based incentives may positively affect the motivation and productivity of the non-profits' managers and employees, there are also several reasons as to why they may not be effective, or even backfire.

⁴ Relatedly, recent studies in the context of for-profits in high-income countries have shown that employees across various occupations (e.g., consultants, lawyers, and online workers) are willing to forgo financial compensation for the pursuit of "meaningful" work (e.g., Bode and Singh 2018, Burbano 2016, Carnahan, Kryscynski, and Olson 2017, Cassar 2019, Cassar and Meier 2018).

⁵ In a similar spirit, non-financial benefits awarded to employees (e.g., through employer recognition, social visibility, or employee satisfaction programs) are found to motivate employees and improve their productivity (e.g., Ashraf, Bandiera, and Jack 2014, Ashraf, Bandiera, and Lee 2014, Dur, Non, and Roelfsema 2010, Flammer and Luo 2017, Gallus and Frey 2016, Gubler, Larkin, and Pierce 2017).

First, employees of non-profits are likely to be purpose-driven and intrinsically motivated. In this regard, the provision of performance-based incentives could crowd out their intrinsic motivation and ultimately reduce their engagement and job satisfaction.⁶ In addition to decreasing employees' intrinsic motivation, such incentives may reduce performance if the employees lack the necessary knowledge on how to improve (Huillery and Seban 2019). Second, compared to financial performance metrics, non-financial performance metrics are difficult to measure and hence likely less suitable as a basis for incentive schemes.⁷ Moreover, the difficulty of measuring, tracking, and verifying non-financial performance might increase the risk of manipulating performance records to increase payouts (Linden and Shastri 2012). Third, in the presence of multi-tasking activities, performance incentives may encourage substitution away from non-incentivized activities, reorienting the employees' attention toward those activities with a greater pay-for-performance component or/and those that require lower effort (Basinga et al. 2011, Holmstrom and Milgrom 1991).

Auditing and feedback (A&F)

In addition to social performance-based incentives, A&F can be introduced to enhance the non-profits' operating efficiency and social performance. Specifically, by conducting audits on a regular basis, independent third parties can verify that best practices and protocols are followed, adequate priorities are set in terms of resource allocation, a strategic and managerial plan is developed and pursued, and that the organization's practices and performance are correctly documented.⁸

Through such audits, the non-profits also obtain data-based feedback on discrepancies between current and target performance. The recognition of underperformance may encourage them to take action

⁶ This argument echoes the findings of studies in the context of for-profit organizations and the provision of financial performance-based incentives (e.g., Bowles 2016, Cassar and Meier 2021, Gubler, Larkin, and Pierce 2016, Wrzesniewski et al. 2014).

⁷ In the for-profit context, pay-for-social performance incentive schemes have been shown to be ineffective when they are vague and insufficiently specified (Flammer, Hong, and Minor 2019).

⁸ For example, in the context of the DRC's health centers examined in this study, regular visits of independent auditors can ensure that general guidelines provided by the WHO are followed (PDSS 2016).

and improve their practices. Moreover, given their business expertise, auditors can play an important role in providing feedback and recommendations to the non-profits' managers and employees, thereby acting as informal coaches.⁹ Such coaching might be of foremost importance. Indeed, an inherent challenge faced by non-profits in low-income countries is to find competent managers and well-trained personnel. The lack of know-how can lead to, e.g., unrealistic planning, inefficient work deployment, inadequate priorities in resource allocation, improper interpretation and implementation of guidelines and procedures, and overall poor decision-making; all of which are likely to result in the inefficient use of the non-profit's already limited financial resources, inventory, and human capital. Accordingly, by sharing valuable insights, auditors can help transfer best practices and improve the non-profits' effectiveness in terms of both their operating efficiency and quality of services. Moreover, feedback can be reassuring for individuals and help them build up their self-confidence that they are doing things right (Bandiera et al. 2015).

On the other hand, A&F may also have detrimental effects on employees. In particular, rating employees may reduce their self-esteem and create animosity and competition among peers (Smither et al. 2005). Relatedly, managers' evaluation of subordinates (which is commonly done in traditional performance evaluation and in "rank-and-yank" performance appraisals) has been found to crowd out creativity, as employees can hesitate to suggest unusual, creative ideas for fear of a harsh evaluation (Amabile et al. 1990).

Financial subsidies

In addition to governance, financial subsidies can play an important role in enhancing the social impact of non-profits, especially in resource-constrained environments. The provision of financial subsidies helps alleviate financing constraints and allows non-profits to invest in, e.g., upgrading their equipment, hiring more employees, and extending the scope of their services. This likely increases the scale of the non-profits' operations. Thus, the provision of financial subsidies in combination with the implementation of

⁹ For instance, auditors of the DRC's health centers examined in this study provide feedback to the audited health centers in order to improve their operations and help them follow best practices (PDSS 2016).

governance mechanisms may lead to improvements in both i) the non-profit's scale of operations, as well as ii) the non-profit's operating efficiency and social performance. As such, financial subsidies and governance mechanisms might serve as complements toward the objective of increasing the non-profit's overall social impact.

That being said, it is unclear whether financial subsidies alone (i.e., without improving the non-profit's governance and management practices) are sufficient to help improve operating efficiency and the quality of services. Indeed, given the challenges and unique nature of non-profits, the sole provision of financial subsidies need not yield improvements along these dimensions as funding per se does not foster knowledge transfer and competence building, nor does it provide incentives to improve the organization's practices.

In the following, we explore the role of governance in improving non-profits' operating efficiency and their social impact in the context of a randomized governance program conducted in the DRC's healthcare sector (Sections 3-5). We also examine the interplay between governance and financial subsidies in the context of this program (Section 6).

3. Program and experimental design

In this section, we describe the PDSS program and experimental design. This section draws from the PDSS manual (PDSS 2016) that contains detailed information about the program.

3.1 The PDSS program

The PDSS program, known formally as “Projet de Développement du Système de Santé” (officially translated by the World Bank as “Health System Strengthening for Better Maternal and Child Health Results Project”) is a five-year nationwide public health program that was launched by the government of the DRC with the aim to improve the utilization and quality of maternal and infant healthcare services in targeted health zones in the country's territory. The program began in 2017 in 156 health districts in 11 of the country's 26 provinces: Kwango, Kwilu, Mai-Ndombe, Équateur, Mongala, Sud-Ubangi, Tshuapa, Haut-Katanga, Haut-Lomami, Lualaba, and Maniema. The program cost of \$521 million is financed by the

DRC government, the World Bank, UNICEF, GAVI, UNDP, and donor country support.

The PDSS program attempts to improve the quantity and quality of primary healthcare services at the health facility level through two main mechanisms. First, the program presents health facilities with financial incentives to provide a “paquet minimum d’activités” (“minimum package of activities”). Defined by the World Health Organization (WHO), these packages put a strong emphasis on maternal and child health. Second, the program presents health facilities with a summary of their performance on a quarterly basis. This information is collected by the contracting and verification agencies (“établissements d’utilité publique,” EUPs) established at the provincial level. These are semi-autonomous entities appointed by the DRC’s Ministry of Health and Ministry of Finance to manage the contracting and procurement of health services with the health facilities. Concretely, they monitor and verify the health performance outcomes of each facility and provide feedback on the facility’s quarterly management plans.

Prior to the PDSS program, other performance-based financing schemes were piloted in the DRC. These schemes presented a select number of health facilities with different rewards, initially taking the form of generic drugs only (2007-2009), thereafter generic drugs and cash (2009-2010), and finally cash only (2011-2017). Further, they were implemented initially in South Kivu only, later in North Kivu, Orientale, Kasai-Occidental and Kasai-Oriental, and lastly in Katanga. Notice that these initial pilots with performance-based financing were undertaken in only a limited number of health zones, which were different from the zones targeted by the PDSS program. Aside from these pilot schemes, we are not aware of major reforms in healthcare that coincided with our period of study, nor of other programs that may have differentially affected our treated health centers and thus biased our estimates.

Assessments of these pilot schemes were at best mixed. Mayaka et al. (2011) evaluated the first two pilot schemes using a qualitative research approach. They highlight the complexity of the schemes, a lack of shared understanding by key stakeholders, a lack of community engagement, and public sector mismanagement, especially corruption, as the main reasons for why the pilots led to mediocre outcomes. Huillery and Seban (2019) examined the impact of the fee (cash only)-for-performance system on health service utilization and health providers’ motivation in 152 health facilities in Haut Katanga. They find that,

relative to the fixed payment system, the fee-for-performance mechanism slightly *reduced* service utilization, and argue that the main explanation for this result is health workers' lowered motivation at work due to a lack of understanding of how to best perform.

Guided by these lessons learned, the design of the PDSS program departed from previous performance-based financing schemes in healthcare in the DRC in three important ways. First, to address the lack of awareness and understanding about evidence-based practices, the PDSS program provides not only financial incentives but also auditing and feedback (A&F) to health facilities. Second, to mitigate problems of mismanagement, the control over the contracting and procurement relationships with each health facility was delegated by the government to third-party agencies at the provincial level. Furthermore, the PDSS program also introduced a community verification system, whereby local associations are mandated to check the actual existence of patients indicated in the facilities' registries and collect information about patients' satisfaction. Taken together, these three distinct design features aimed to improve the accountability of government and health workers, and the alignment of interests between the health system's key stakeholders: health workers, funders of the PDSS program, national and provincial governments, and patient communities.

The background context in which the PDSS program was conceived is one marked not just by poor health outcomes but also serious health system challenges. The DRC's human development indicators are among the lowest in the world, and four decades of conflict and mismanagement severely weakened the country's institutions and infrastructure. Moreover, the DRC is not on track to achieve any of the Millennium Development Goals (MDGs), including those related to maternal and child health. The main maternal and child health indicators remain very poor. According to UNICEF (2016), the maternal mortality ratio is 693 (per 100,000 live births) and the under-five child mortality rate is 98 (per 1,000 live births). Decreasing these mortality rates requires improvements in both the quantity and quality of reproductive and child health services. Together, these challenges help explain why the PDSS program is designed the way it is, and in particular why the financial incentives provided by the PDSS program are primarily linked to maternal and child health outcomes.

3.2 Performance incentives

The PDSS program presents health facilities with financial incentives based on performance indicators measured at the health facility level (as opposed to the level of individual health workers). The size of the financial transfers (T) that health facilities receive depends on the performance along 22 health indicators and overall service quality. The transfers paid to the health facilities are mainly used for purchasing essential equipment. No more than fifty percent can be used toward health workers' salaries. Payments are made directly to the health facilities.

The overall transfer is comprised of two parts:

$$T = \sum_{i=1}^{22} p_{is}q_i + \alpha$$

where the first part of the transfer reflects the performance along the 22 healthcare performance indicators listed in Table A1. For each indicator, a target is set depending on the size of the population served by the health facility. The payment amount per unit of health service i provided, p_{is} , varies depending on the percentage share s of the target performance achieved. For each indicator i , the closer the actual performance is to the targeted performance, the higher the unit price. Some of these indicators refer to the reasons for a visit (such as prenatal care or delivery), whereas others refer to the type of services provided during a visit (such as tetanus vaccination during prenatal care). All the indicators and payment amounts are defined in concordance with WHO guidelines.

The second part of the transfer reflects the overall health service quality. The parameter α denotes the “quality bonus,” a supplement of up to 25% of $\sum_{i=1}^{22} p_{is}q_i$. Quality is assessed using a quality assessment grid that consists of a longlist of evidence-based practices. The grid considers practices in 15 distinct domains, ranging from the overall organization and management of the health facility to practices in terms of HIV/tuberculosis, and hygiene and sterilization. The list is provided in Table A2. A total of 400 points can be earned. To receive a bonus, the health center must secure a minimum of 200 points, that is, obtain a quality score of at least 50 percent. When the quality score is above 50 but below 80 percent, the quality

bonus is given by $\alpha = \text{quality score} \times 25\% \times \sum_{i=1}^{22} p_{is} q_i$. When the quality score is above 80 percent, then the quality bonus is given by $\alpha = 25\% \times \sum_{i=1}^{22} p_{is} q_i$.

Health facilities submit monthly activity reports and quarterly requests for payment to the EUPs, who are responsible for the verification of data and authorization of payments. A team of auditors verifies the reported number of medical acts in the minimum package of activities (MPA) delivered by the center during the period, and reviews the quality of the services provided using the quality assessment grid. The auditors are recruited by and are accountable to the EUPs. They are helped by local associations whose role is to question the community to i) find out whether the patients declared by the centers have effectively received the corresponding treatments, and ii) gather feedback regarding their satisfaction. The team of auditors not only assesses the health facilities' performance outcomes and practices, it also provides feedback and recommendations. The grid provides objective data regarding discrepancies between current practices and target performance. Every quarter, the health facilities are expected to draw on these inputs to plan realistic and progressive improvements. The auditors support the write-up of the quarterly management plans that define how the improvements can be achieved. They also provide training in finances and stock management.¹⁰

3.3 Selection into the program and randomization

The PDSS program was implemented in 11 of the DRC's 26 provinces. The 11 provinces were selected based on three criteria: poor health indicators, limited access to health services, and the ability to build on or expand an ongoing partnership with an international organization. Only health centers registered with the DRC Ministry of Health were eligible for the program. (They need not be affiliated with the government, though.) Within each province, the selection of health facilities was made by Médecins d'Afrique (MDA), an international NGO, in close collaboration with the World Bank's impact assessment team and the PDSS

¹⁰ The auditors are members of the World Bank's Development Policy Staff (DPS) who were previously deployed in the DRC and have expertise in healthcare. On top of their medical training, they received additional training by World Bank specialists in performance-based financing. Note that it is usually the same team of auditors who visit a given health center each quarter. This is meant to ensure a certain continuity in the feedback that is provided. A given team of auditors is often affected to several health zones within a given province.

project unit. Together, they conducted a baseline evaluation for which they designed questionnaires and protocols for establishing the selection. They then assessed the health centers' suitability for the program as health centers were required to have a minimum level of quantity and quality of services. The baseline evaluation was conducted between June 2015 and March 2016. The selected health centers were then randomly assigned into a treatment group ("Groupe cas") and control group ("Groupe témoin"). The randomization itself was done in Excel, and every third health center was assigned into the control group.

It is important to note that both the treated and control health centers are part of the PDSS program, in that they both receive funding from the program. The payments are twofold. First, at the beginning of the program (first quarter of 2017), health centers in both the treatment and control groups receive an initial subsidy (called "unité d'investissement") that is meant to finance the purchase of essential equipment. Second, in the first month of each quarter (starting in the second quarter of 2017), the performance-based subsidy is paid out to the treated health centers based on the auditors' assessment conducted in the previous quarter (the formula used to compute the subsidy is described above in Section 3.2). Control health centers also receive a quarterly subsidy that matches the average subsidy received by the treated health centers in the same quarter. As such, the quarterly payments are on average the same in both the treatment and control groups.

By design, the PDSS program provides an ideal setup for our study. By comparing health centers that are randomly assigned to the treatment versus control group, we are able to identify the impact of the governance intervention—that is the provision of performance-based incentives combined with A&F—on health centers' operating efficiency and social performance, holding everything else, *including funding*, constant.

4. Data and methodology

4.1 Data sources

The data on the DRC health centers are obtained from administrative records of the DRC Ministry of Health that are maintained in the DRC's Système National d'Information Sanitaire (SNIS). They include detailed

information on the health centers' operations (e.g., staff, number of consultations, number of births), along with the name and location of each health center.

These data were supplemented with identifiers for the treated and control health centers provided by the World Bank. Our baseline sample includes a total of 999 health centers, out of which 674 were assigned to the treatment group, and 325 to the control group. For each health center, we were granted access to 14 quarters of data, ranging from the first quarter of 2016 until the second quarter of 2019. That is, our dataset includes four quarters before and ten quarters after the program's launch.

Note that, while the large-scale implementation of the PDSS program took place in the first quarter of 2017, a small number of health centers were already treated in 2016 in a pilot-like setting. Those are excluded from our sample.

4.2 Outcome variables

Our objective is to study how the governance intervention—that is, the provision of performance-based incentives combined with A&F—affects health centers' outcomes. In what follows, we describe the outcome variables.

Health center's operating efficiency

We compute a health center's operating efficiency as the number of primary healthcare services performed divided by the number of employees. This measure captures the health center's labor productivity (i.e., output per employee). Note that the health centers in our sample only offer primary healthcare services. Secondary healthcare services are typically administered at hospitals, often upon referrals from the health centers.

Health center's employees

We use several variables to examine changes in the health center's staff. First, we use the total number of employees working at the health center. Second, we decompose this total into the number of doctors, nurses, and administrative personnel, respectively.

Volume of healthcare services

To measure the volume (or “quantity”) of healthcare services, we use the number of primary healthcare services performed. Since antenatal care and childbirth are the main services performed at the health centers, we also use two additional metrics: the number of maternity and childhood healthcare services performed, and the number of births.

Quality of healthcare services

To measure the quality of healthcare services, we focus on infant mortality at birth. This is a key metric in our context since antenatal care and childbirth are the main services performed at the health centers. Infant mortality at birth can occur in two forms: stillbirth and neonatal death. Stillbirth refers to a baby born with no sign of life at or after 28 weeks of gestation; neonatal death refers to a baby who dies within the first 28 days of life (WHO 2019a, 2019b). We compute the ratio of stillborn babies to the total number of births (henceforth “share of stillbirths”), the ratio of neonatal deaths to the total number of births (“share of neonatal deaths”), and the complement (“share of live births”). These three ratios allow us to assess not only the quality of the childbirth services per se, but also the quality of antenatal care services. Indeed, medical research has shown that antenatal care reduces the likelihood of stillbirth and neonatal death (Adam et al. 2005, Hollowell et al. 2011). As such, the above measures capture the quality of the main services performed at the health centers.

4.3 Summary statistics

Table 1 provides summary statistics for the 999 health centers in our sample. All statistics refer to the fourth quarter of 2016 (i.e., the quarter that precedes the start of the PDSS program). On average, health centers in our sample performed 1,611 primary healthcare services (236 on a per employee basis). The majority (957 out of 1,611) were maternity and childhood healthcare services. As discussed above, antenatal care and childbirth are the main services performed at the health centers; the summary statistics reflect this institutional feature of the DRC’s healthcare system.

----- Insert Table 1 about here ----

Other statistics are worth highlighting. The average number of employees is 7.1, consisting mainly of nurses and administrative personnel. The average (quarterly) number of births is 53.8, out of which 0.70% are stillbirths, and 0.17% neonatal deaths. Lastly, as can be seen from the bottom panel, the average health center is located in a health district of 204,409 inhabitants, and a health area of 11,135 inhabitants.¹¹ To further characterize the health centers from our sample, Figure 1 provides photographs featuring three of them.

----- Insert Figure 1 about here ----

4.4 Randomization tests

Our identification strategy relies on the random assignment of health centers to the treatment and control groups. Since randomization is a feature of the PDSS program, this requirement should hold by design.

To empirically assess the validity of the randomization, we can examine the covariate balance prior to the treatment—intuitively, if the assignment is truly random, there should be no systematic difference between health centers in the treatment and control groups based on pre-treatment characteristics.

We conduct this analysis in Table 2, where we report the same set of summary statistics as in Table 1, but separately for the 674 health centers in the treatment group and the 325 health centers in the control group. The statistics are again computed in the fourth quarter of 2016 (i.e., the quarter that precedes the launch of the PDSS program). The last two columns of the table provide the p -value of the difference-in-means test and the Kolmogorov-Smirnov (KS) test of identical distributions, respectively, for each covariate.

----- Insert Table 2 about here ----

As can be seen, there is no systematic difference between the treated and control health centers. For all covariates, the summary statistics are very similar in economic terms. They are similar in statistical

¹¹ The DRC is partitioned into 516 health districts (also called health “zones”) and 8,504 health areas. Appendix A provides a description of the DRC’s health system along with a characterization of the health districts and health areas, respectively.

terms as well. Specifically, the difference-in-means test (reported in the penultimate column) is always insignificant, with p -values ranging from 0.196 to 0.981. Similarly, the KS test of identical distributions (reported in the last column) is always insignificant as well, with p -values ranging from 0.140 to 0.980. Overall, the evidence from Table 2 confirms the random assignment of health centers.¹²

This randomization is further illustrated in Figure A3, where we plot the distribution of each outcome variable in the treatment vs. control groups in the quarter preceding the treatment. As can be seen, the distributions are very similar for each variable. The formal test for identical distributions—the KS test from Table 2 mentioned above—confirms that the null of identical distributions cannot be rejected for all variables.

4.5 Methodology

To examine how the treatment affects health center outcomes, we use a difference-in-differences methodology. For each health center and each outcome variable y (e.g., share of stillbirths, share of neonatal deaths, operating efficiency), we compute the change from $t - 1$ (the quarter preceding the intervention, Q4 2016) until $t + 9$ (nine quarters after the intervention, Q2 2019), which we denote by $\Delta y_{t-1,t+9}$. Note that the quarter of the treatment is denoted by $t = 0$ (Q1 2017), and hence the post-treatment period consists of 10 quarters from $t = 0$ until $t = 9$. Whenever y is a ratio (e.g., the share of stillbirths), Δy represents the difference in the ratio; whenever y is a level (e.g., the number of employees), Δy represents the percentage change (denoted by $\% \Delta y$ in the tables). We then estimate the following difference-in-differences specification:

$$\Delta y_{i,t-1,t+9} = \alpha_p + \beta \times treatment_i + \varepsilon_i \quad (1)$$

where i denotes health centers; α_p are province fixed effects; $treatment$ is the treatment indicator, which is equal to one for health centers in the treatment group (and zero for those in the control group); and ε is the

¹² In Figure A2, we plot the location of the health centers in the control group (blue markers) and treatment group (green markers) on the DRC map. We caution that the map is incomplete as granular geo-codes (and hence the longitude-latitude coordinates) are only available for about 73% of the sample. As can be seen, there is no apparent imbalance between the two groups. More formally, when we examine the distribution of the control and treated health centers within each province, we find no significant deviation from the program's targeted 2/3 of treated health centers. Specifically, the null of a 2/3 distribution within each province cannot be rejected with a p -value of 0.268.

error term. We cluster standard errors at the health district level to account for potential dependence of the error term at the local level. The coefficient of interest is β , which measures the difference in $\Delta y_{t-1,t+9}$ between the treated and control health centers (i.e., the difference-in-differences). In other words, it measures the effect of the governance treatment on y accounting for contemporaneous changes in y at the control health centers. Note that the inclusion of province fixed effects ensures that we compare treated and control health centers within the same province. (In robustness tests, we show that our results hold regardless of their inclusion.)

In our main analyses, we also estimate variants of equation (1) to examine the dynamics. Specifically, we consider $\Delta y_{t-4,t-1}$, $\Delta y_{t-3,t-1}$, and $\Delta y_{t-2,t-1}$, to test for pre-trends, and $\Delta y_{t-1,t}$, $\Delta y_{t-1,t+1}$, ..., $\Delta y_{t-1,t+9}$ to characterize the dynamics of the treatment effect.

5. Results

5.1 Impact of the governance treatment on health center outcomes

Main results

Table 3 presents our main results. The estimates are obtained from regression (1) using the 999 health centers in our sample. In columns (1) and (9)-(11), where the dependent variable y is a ratio, Δy represents the difference in y from $t - 1$ to $t + 9$. In columns (2)-(8), where the dependent variable y is a level, $\% \Delta y$ represents the percentage change in y from $t - 1$ to $t + 9$.

----- Insert Table 3 about here -----

As can be seen from column (1), operating efficiency increases significantly following the governance treatment. The coefficient of 101.4 (p -value = 0.000) implies that the number of primary healthcare services per employee increases by 101.4. Given a pre-treatment average of 238.5 (Table 2), this corresponds to a 42.5% increase in operating efficiency. This finding indicates that the governance intervention leads to higher operating efficiency.

In columns (2)-(8), we unpack this change in operating efficiency by examining changes in the numerator (number of services performed) and denominator (number of employees). In terms of the

denominator, we find that, following the treatment, health centers significantly reduce their administrative staff—the number of administrative employees decreases by 9.6% (p -value = 0.085)—while there is virtually no change in the number of doctors and nurses. In terms of the numerator, we find that the level of services performed increases, but not significantly so. Overall, this finer analysis suggests that the efficiency gains are derived mainly from a reduction in administrative overhead.

Note that the decrease in administrative overhead is not about layoffs per se. In fact, both the treatment and control group increase their staff from $t - 1$ to $t + 9$. The observed difference between the two groups indicates that treated health centers hire fewer additional administrative employees compared to the control group.

In columns (9)-(11), we find that the quality of healthcare services increases significantly following the treatment. Specifically, we find that the share of stillbirths decreases by 0.32 percentage points (p -value = 0.006), and the share of neonatal deaths by 0.23 percentage points (p -value = 0.032). Correspondingly, the share of live births increases by 0.55 percentage points (p -value = 0.002). Put differently, for every 1,000 new births, the governance treatment helped save about 3.2 lives at birth (reduction in stillbirths) and 2.3 lives within the first 28 days after birth (reduction in neonatal deaths). These are large effects in light of the baseline probabilities. Indeed, the pre-treatment share of stillbirths and neonatal deaths is on average 0.91 percentage points (Table 2). Hence, a decrease by 0.55 percentage points in the share of stillbirths and neonatal deaths corresponds to a 60.4% reduction in infant mortality. Overall, these results indicate that the governance intervention brought about large improvements in social performance.¹³

Dynamics

In Table 4, we examine the dynamics of the treatment. To do so, we re-estimate regression (1), but instead of computing changes in outcomes from $t - 1$ to $t + 9$, we compute changes in outcomes for increasing time

¹³ In Appendix B, we present several robustness checks. Specifically, we show that our results are robust if we i) do not include province fixed effects (Table A3), ii) do not cluster standard errors (Table A4), and iii) re-estimate the regressions of the share of stillbirths, neonatal deaths, and live births using weighted least squares (WLS), weighting observations by the number of births, thereby accounting for the fact that ratios of births are more accurately measured when the number of births is higher (Table A5).

intervals on a quarterly basis ($\Delta y_{t-1,t}, \Delta y_{t-1,t+1}, \dots, \Delta y_{t-1,t+9}$) and also consider pre-trends ($\Delta y_{t-4,t-1}, \Delta y_{t-3,t-1}$, and $\Delta y_{t-2,t-1}$). Each estimate in the table is obtained from a different regression, depending on the outcome variable (columns) and time horizon (rows). In Figure 2, we plot the coefficients in event time (along with 90% confidence bounds) for each outcome variable.

----- Insert Table 4 and Figure 2 about here -----

As can be seen, we find no evidence for pre-trends; the corresponding point estimates are all small and insignificant. We also find no evidence for an immediate response at the time of the intervention. Rather, the benefits from the treatment take a few quarters to materialize. The first tangible effect is observed after four quarters, when the increase in operating efficiency (and the underlying reduction in administrative overhead) becomes significant. The improvements in the quality of healthcare services take even longer to materialize. It is only after seven quarters that the birth mortality statistics start showing significant improvements.

The delayed response is consistent with the gradual adoption and learning of better practices. This was confirmed in a series of interviews we conducted with program participants ($N = 20$).¹⁴ Collectively, the respondents highlighted that a key challenge was the lack of adequate training and organizational know-how. In this regard, the quarterly rounds of feedback (combined with incentives to actually act on this feedback) were seen as essential in inducing tangible changes that would ultimately translate into higher efficiency and quality of services.

It is also worth noting that the long learning curve of the health centers' employees was a key consideration in designing the PDSS program. As discussed in Section 3.1, prior to the PDSS program, pilot programs using performance-based financing designs were run in selected parts of the DRC. These programs were unsuccessful, suggesting that the mere use of performance-based financing is insufficient to induce tangible improvements. This led to the innovative design of the PDSS that combines

¹⁴ We conducted 20 interviews with individuals involved in the PDSS program, out of which 17 were based in the DRC. The interviewees included health center employees (nurses and doctors), auditors, and programs administrators. The recordings of these interviews are available upon request.

performance-based incentives with A&F.

Lastly, the longer lag we observe for the improvements in infant mortality (7 quarters, compared to 4 quarters for the improvements in operating efficiency) reflects the type of services provided at the health centers. Indeed, the bulk of the health centers' services pertain to pre-natal care. By their very nature, improvements in pre-natal care reduce the risk of stillbirths and neonatal deaths several months in the future, and are likely more effective when higher-quality services are provided throughout the full pregnancy cycle. This likely explains the longer lag we observe in the data.

Economic magnitudes

As mentioned above, our baseline estimates are large in economic terms—the estimates from Table 3 imply that operating efficiency increases by 43% and infant mortality decreases by 60%. Our interviews of program participants help shed light on these magnitudes. All respondents ($N = 20$) expected the improvements to be large, citing the low education level and inadequate training of the health center employees as key rationales for a steep learning curve. One respondent even qualified the benefits of the program as “énormes” (enormous) due to the above reasons.

While interviews are subjective in nature, a perhaps more objective benchmark is provided by the World Bank, who targeted to reach a 65% quality score for the treated health centers, compared to a 20% score prior to the start of the PDSS program (World Bank 2018). This corresponds to more than a three-fold increase in quality.

To further assess the magnitudes, and put them into perspective, we benchmark our estimates against related estimates from the literature. In this regard, the literature on management practices is especially informative, as considerable effort has been put into quantifying the productivity gains from different management practices. (In spirit, our governance treatment consisting of social performance-based incentives combined with A&F can be seen as a form of “structured management practice” in the terminology of Bloom et al. 2019). Specifically, in their study of management practices at U.S. manufacturing plants, Bloom et al. (2019) find that a one standard deviation increase in their score of

structured management is associated with a 26% percent increase in labor productivity, noting that “[t]he magnitude of the productivity-management relationship is large” (p. 1649). Giorcelli (2019) finds that Italian companies who participated in a management assistance program achieved a 49% increase in productivity. Similarly, Bruhn et al. (2017) find that Mexican small and medium enterprises that were offered a one-year access to management consulting services achieved a 27% increase in productivity.

Naturally, we caution that the benefits of the improved management practices considered in these articles need not be directly comparable to those of the intervention considered in our study given the different nature of non-profit organizations. Nevertheless, they do provide guidance in terms of how changes in business practices can affect operating efficiency. Collectively, they indicate that the efficiency gains can be substantial, in the ballpark of what we find in our setting. In addition, such efficiency gains might be larger in developing countries such as the DRC due to the potentially bigger room for improvement, as several respondents pointed out in our interviews.¹⁵

Post-treatment distributions

In Figure A4, we plot the post-treatment distributions of each outcome variable in the treatment and control groups, respectively. In principle, the intervention could lead to a separation of the treated health centers into “superstar” health centers (i.e., those that do well based on the performance targets) and “failure” health centers (i.e., those that do not). In this scenario, the tails of the performance improvements would be wider in the treatment vs. control group. In Panel A of the figure, we plot the post-treatment distributions of operating efficiency in the treatment and control groups. As can be seen, we find that the right tail improves in the treatment (vs. control) group. However, we do not find evidence for a deterioration in the left tail. In

¹⁵ Another way to put the magnitudes into perspective is through the VSL (value of a statistical life) framework. Viscusi and Masterman (2017) estimate a VSL of \$71,000 for the DRC (compared to a VSL of \$9.6M for the U.S.). Our baseline results in Table 3 suggest that the governance treatment reduces infant mortality by 0.55%. Factoring in the number of treated health centers (674) and the average number of births per health center (55.7, see Table 2), the governance intervention corresponds to VSL gains in the amount of \$14.7M. Compared to the cost of the program (\$521M), this translates into a VSL-ROI of about 3% from governance alone. Naturally, we caution that this number likely represents a lower bound since the improvements brought about by the governance intervention are likely to improve patient health in ways that are not captured by our infant mortality metrics.

fact, the left tail improves as well. More broadly, the whole distribution seems to shift to the right (compared to the control group). In Panel K, we observe a similar pattern when looking at the post-treatment distributions of the share of live births. (A consistent pattern is also observed in Panels I and J when looking at the share of stillbirths and share of neonatal deaths, respectively.) Overall, this suggests that the treated health centers that score low in the quarterly assessments (and hence receive less funding than the control health centers) do not perform worse than the control group.

This finding is consistent with the World Bank’s insights from their earlier pilot programs in which they found that the sole provision of performance-based funding did not affect the performance of health centers in the RDC. This is also consistent with our auxiliary findings based on the “outside” group, where we find that funding per se is not associated with subsequent changes in performance (see Section 6). Rather, this finding highlights the importance of combining performance-based incentives with quarterly rounds of auditing and feedback. By implementing at least some of the feedback—which is likely to be the case for most treated health centers—performance improves. In contrast, not implementing the feedback is conducive to the status quo and hence need not hurt performance (compared to the counterfactual provided by the control group).

5.2 Potential challenges and alternative interpretations

In this subsection, we discuss potential challenges of the PDSS experiment, and provide additional evidence that mitigates alternative interpretations of our findings.

Contamination

An important concern is that the treatment might be contaminated by strategic interactions among nearby health facilities. In particular, health centers may try to “game” the performance-based incentives by strategically refusing high-risk patients (e.g., pregnant mothers who face a high risk of stillbirth), referring them to nearby health centers and hospitals. Such gaming behavior could potentially explain our results.

Nevertheless, this concern is mitigated, for two reasons. First, as part of their assessment protocol, the auditors conduct interviews among the local community. Accordingly, if health centers were to transfer

high-risk patients elsewhere, they would likely find out. Second, in Panel A of Table 5, we examine whether the share of stillbirths and neonatal deaths increases (and the share of live births decreases, respectively) at other healthcare facilities that are located in the same health district as the treated health centers. Those include hospitals (columns (1)-(3)), control health centers (columns (4)-(6)), and health centers not in the PDSS program (columns (7)-(9)) that are located in the same health district as the treated health centers. As can be seen, we find that all point estimates are small and insignificant, which is inconsistent with the gaming interpretation. In Panel B of Table 5, we further examine whether the number of services changes at the nearby healthcare facilities (number of primary healthcare services, number of maternal and childhood healthcare services, and number of births). Again, we find no such evidence.

----- Insert Table 5 about here -----

A related concern is that the treated health centers may expand their capacity post-treatment, and hence be able to see a higher number of low-risk patients who would have otherwise relied on home care. In this scenario, the treated health centers would have a higher share of low-risk patients, which could explain the improvements in the birth statistics. Nevertheless, as shown in column (8) of Tables 3 and 4, the increase in the number of deliveries is not significantly higher in the treatment vs. control group. This mitigates the possibility that the improvements in infant mortality are purely driven by “scale” (and the fact that higher scale may help clinics admit a large share of low-risk patients). Moreover, in our interviews, it was noted that the choice to deliver at home or at the health center is not made based on the risk assessment of the pregnancy, as women themselves cannot easily assess the level of risk of their pregnancy. Rather, the stated rationales were the perceived quality of services offered by the health centers and the value for money of these services. Neither is related to the pregnancy’s riskiness.

Corruption

Another potential challenge is that the health center staff may attempt to bribe the auditors. To alleviate this possibility, the PDSS program has explicit protocols to prevent and sanction fraud. Specifically, the program collaborates with an external counter-audit agency that periodically audits the auditors and helps

guarantee the integrity of the process. In case of a discrepancy between the evaluation of the auditors and the evaluation of the counter-audit agency, and if the discrepancy exceeds 10% of the composite score (and no qualitative justification is provided to explain the difference), both the auditor and the health center are sanctioned. For auditors, the first offense triggers the retention of 30% of the auditor team's compensation, and the second offense triggers the suspension of the auditor contract until administrative measures are taken. For health centers, the first penalty is a retention of 30% of the total subsidies from the next quarterly payment, and the second penalty is the termination of the performance contract until administrative measures are taken (such as the replacement of the head of the health center by a candidate whose application is approved by the Ministry of Health).

While these protocols alleviate the possibility of misreporting, it is worth noting that—even if misreporting occurs—it is unlikely to contaminate our analysis. Indeed, our data on health center outcomes are obtained from the DRC Ministry of Health (not the World Bank). Accordingly, we do not examine the specific performance indicators used by the World Bank, but instead study performance outcomes (such as the number of stillbirths, neonatal deaths, etc.) that are collected independently of the PDSS program.

Employee turnover

Our results show that the treated health centers hire fewer administrative employees post-treatment, while there is no significant change in the hiring of nurses and doctors. That being said, even if we observe no significant change in a given employee category, it could be that low-skill employees were replaced by high-skill employees. In this scenario, our results could be driven by the recruitment of higher-skilled employees, as opposed to the governance treatment itself. Unfortunately, we do not have disaggregated data at the employee level that could be used to examine this alternative. Nevertheless, we believe it is unlikely to explain our results, for the following reasons.

First, the labor market for health workers is very thin in the DRC. The WHO estimates that the DRC only has 6 qualified health workers per 10,000 population, as compared to the WHO's recommended minimum of 23 per 10,000 population (WHO 2016). Second, and importantly, the mobility of health

workers is highly restricted, as their allocation is centralized at the province level. The entry-level market relies on the initial training of medical staff that is exclusively done in the major cities and abroad. Periodically, the National Ministry of Health allocates new health workers to each province, and the respective Provincial Ministry of Health then allocates them to health centers within the province. The reallocation of health workers across health centers is also handled by the respective Provincial Ministry of Health. In other words, each Provincial Ministry of Health has full decision rights over the recruitment and distribution of medical staff across health centers.

Accordingly, labor mobility is not under the direct control of the health centers nor the health workers. What health centers can do is sanction or request the mobility of a specific health worker. The request is then handled at the provincial level. Similarly, health centers can request an increase in their medical staff, but have little influence on the selection process of the personnel they receive. For this reason, it is unlikely that the treated health centers will be able to hire “better” medical staff post-intervention.¹⁶

Hawthorne effect

A potential concern with randomized controlled trials is the so-called “Hawthorne effect”—that is, the possibility that the treated subjects might feel observed and alter their behavior in response (for reasons unrelated to the treatment). In particular, it could be that being in the PDSS program induces a status boost that could affect the staff’s behavior regardless of the intervention per se. Nevertheless, this concern is mitigated, for two reasons. First, control health centers were an integral part of the PDSS program, as they received a significant amount of subsidies from the program (the same average amount as the treated health centers). Our qualitative interviews further confirmed that the control health centers clearly felt part of the PDSS program. Second, a pure status boost would likely affect outcomes already in the early quarters of the program, while we find that the effects materialize with a lag of several quarters.

¹⁶ In contrast to the medical staff, non-medical staff (that is, administrative employees) are recruited by the health center board. As such, their number can be optimized at the health center level. Nevertheless, the higher mobility of administrative employees is unlikely to explain the higher quality of medical services per se, as administrative employees are barred from performing medical acts by the PDSS guidelines.

Implementation challenges

In our interviews of program participants, two implementation challenges were mentioned. First, a non-trivial challenge was accessibility. Some villages and health facilities were harder to access due to poor road conditions, locations in flooded riverine areas, or because of security problems. In several instances, auditors had to use canoes and motorcycles. The second challenge was connectivity. Although auditors were equipped with cell phones and tablets, some teams were out of range of communication systems for days. This limited their ability to take full advantage of the features of the platforms used to enable online data collection and retrieval. These logistical challenges led to occasional delays in conducting the auditing. That being said, given their idiosyncratic nature, they are unlikely to substantially affect our analysis.¹⁷

5.3 Cross-sectional heterogeneity

The results presented so far indicate that the governance intervention—i.e., the provision of performance-based incentives combined with A&F—brings about large improvements in operating and social performance. As discussed above, these findings point toward a “learning” interpretation. In particular, the fact that the effects materialize after a few quarters (as opposed to the very short run) is consistent with the health center staff progressively learning from the quarterly rounds of feedback, and having incentives to act on this feedback.

To shed further light on the mechanism, we examine how our results differ depending on various cross-sectional characteristics. The first variable we consider is the density of health centers in the health district, measured as the number of health centers per capita (i.e., the number of health centers divided by the health district’s population). In low-density areas, health centers have limited opportunities to learn and adopt best practices from their peers. Accordingly—to the extent that the auditors’ quarterly rounds of feedback help transfer useful knowledge and best practices—the treatment effect might be stronger among those.

¹⁷ Another potential challenge was the Ebola outbreak of 2014-2016 that preceded the launch of the PDSS program. While the Ebola crisis most certainly affected the health centers in our sample, it is unlikely to significantly distort our estimates as it affected both the treatment and control health centers.

The second set of variables we consider are measures of ex ante performance, namely i) operating efficiency and ii) infant mortality (the sum of stillbirths and neonatal deaths, divided by the number of births). Low-performing health centers are more likely to lack proper know-how and hence might be more responsive to the treatment. Finally, we also consider the moderating role of health center size, as smaller health centers might have a lower knowledge base to begin with.

In Table 6, we augment regression (1) by interacting *treatment* with the four measures described above. Note that all interaction terms are measured in the quarter that precedes the intervention (that is, at $t - 1$), except for infant mortality which is measured over the four quarters prior to the treatment (that is, from $t - 4$ to $t - 1$). We use four quarters to ensure that infant mortality has enough cross-sectional variation, as infant mortality is often 0 for a given health center in a given quarter. For ease of interpretation, we standardize each interaction term. That is, the coefficients capture the differential treatment effect for a one-standard deviation increase in the variable of interest. Note that the regressions also include the interacted variables as standalone controls.

----- Insert Table 6 about here -----

As can be seen, we find that the local density of health centers significantly moderates our results. Specifically, we find that health centers in low-density areas experience a higher increase in operating efficiency and a larger decrease in infant mortality. These findings are consistent with the learning interpretation, as they suggest that health centers benefit more from the treatment when it is harder to learn and adopt best practices from their peers. Moreover, we find that i) the increase in operating efficiency is higher for health centers with lower ex ante operating efficiency, and ii) the decrease in infant mortality is more pronounced for health centers with higher ex ante mortality. These findings indicate that the improvements brought about by the treatment are larger when there is more to improve and learn to begin with, which again points toward the learning interpretation. In contrast, health center size per se does not significantly moderate the treatment effect.

Naturally, we caution that the results in Table 6 are correlational, as we do not have instruments

for the interacted variables.¹⁸ Nevertheless, collectively, these results lend support to the learning interpretation—that is, the health center staff learn from the quarterly rounds of feedback (while being incentivized to act on this feedback), which ultimately translates into higher operating efficiency and social performance.

In Table A6, we refine the analysis of past performance by considering the lower tail of the ex ante performance distribution. Specifically, instead of using continuous measures, we now use two dummy variables that indicate whether the health center is in the bottom 25% and 10%, respectively, of economic performance (measured using operating efficiency; columns (1)-(2)) and social performance (measured using infant mortality; columns (3)-(4)). We find that the treatment effect is significantly stronger among the worse-performing clinics. The increase in operating efficiency is 0.4 (0.5) times larger for clinics in the bottom 25% (10%) of economic performance; the decrease in infant mortality is 1.7 (1.9) times larger for clinics in the bottom 25% (10%) of social performance. This confirms that the effectiveness of the intervention is greater among low-performing clinics. Interestingly, we find that the treatment effect, albeit smaller in economic terms, remains significant even outside the bottom 25% of clinics based on their ex ante performance. Arguably, given the many challenges faced by the healthcare sector in the DRC, there is considerable room for improvement, even leaving aside the lowest-performing clinics.

6. Funding vs. governance: auxiliary evidence from health centers outside the PDSS program

In the analysis presented so far, we focused on health centers that were selected for the PDSS program—that is, health centers that were either in the control group (receiving PDSS funding) or in the treatment group (receiving PDSS funding and the governance treatment).

In this section, we consider a third group of health centers, namely those that are not part of the PDSS program. We refer to this group as the “outside group.” By construction, the outside group does not

¹⁸ In particular, we note that health centers in low-density areas tend to be smaller, have lower operating efficiency, and higher infant mortality. Specifically, density has a correlation of 7.1% (p -value = 0.024) with size, 10.2% (p -value = 0.001) with operating efficiency, and -7.9% (p -value = 0.013) with infant mortality. The inclusion of all four characteristics in the same specification alleviates the possibility that density may merely capture the moderating role of these correlated characteristics.

receive any funding nor the governance intervention from the PDSS program. Accordingly, we can use the outside group to examine the relationship between funding and governance. Specifically, by comparing the treatment group (that receives funding and the governance treatment) versus the outside group (that receives neither), we can assess the benefits of the ‘funding and governance’ bundle. Similarly, by comparing the control group (that receives funding) versus the outside group, we can assess the effectiveness of ‘funding only.’

Our dataset from the DRC Ministry of Health covers a total of 5,832 health centers in the outside group. Table A7 provides summary statistics for those. Compared to the 999 health centers in our sample, the “outside” health centers are on average smaller (5.2 versus 7.1 employees) and provide a lower volume of services (1,153 versus 1,611 primary healthcare services performed). These differences are not surprising. Indeed, as described in Section 3.3, inclusion in the PDSS program is not random. (What is random is the assignment to the treatment versus control group *within* the PDSS program.) Hence, a caveat of using the outside group is that we can no longer rely on randomization, and hence any such analysis is correlational per se.

To mitigate this caveat, we use a nearest-neighbor matching in which health centers in the treatment group (and control group, respectively) are matched to health centers in the outside group based on a large set of observables. Specifically, for each of the 674 health centers in the treatment group (and for each of the 325 health centers in the control group, respectively), we match the nearest—i.e., most similar—health center out of the pool of 5,832 health centers in the outside group. The matching is done in two steps. First, we require that the matched health center be located in the same health district as the treated health center. (If less than 10 outside health centers are available within the relevant district, we relax this requirement and require that the matched health center be located in the same province as the treated health center.) Second, out of the remaining candidates, we select the nearest neighbor based on the health center characteristics in Table A7 (measured in the quarter that precedes the PDSS program). The nearest neighbor is then the one with the lowest Mahalanobis distance to the treated health center along the matching

characteristics.¹⁹

This matching procedure ensures that the matched health centers from the outside group are as similar as possible to the treated health centers (and control health centers, respectively) ex ante. The covariate balance, provided in Table A8, confirms that there is no significant difference between the respective groups.

Treatment group versus outside group: assessing the effectiveness of funding and governance combined

To compare outcomes in the treatment versus outside groups, we estimate a variant of regression (1), using the outside group in lieu of the control group. That is, we now compare health centers in the treatment group versus health centers in the outside group (i.e., health centers that receive neither PDSS funding nor the governance treatment). This comparison allows us to examine how the combination of both funding and governance affects health center outcomes.

The results are provided in Panel A of Table 7. As can be seen, we find again that operating efficiency (column (1)) and the quality of healthcare services (columns (9)-(11)) significantly improve following the treatment. These results mirror those from Table 3. Importantly, compared to the outside group, we find that the treated health centers significantly increase the scale of their operations—the number of employees increases by 13.7% (p -value = 0.001), the number of primary healthcare services by 89.7% (p -value = 0.001), the number of maternal and childhood healthcare services by 92.4% (p -value = 0.001), and the number of births by 69.1% (p -value = 0.000). Arguably, the ‘funding and governance’ bundle benefits the treated health centers in two ways: the additional funding allows them to grow their operations, while the governance intervention helps them improve their economic and social performance.

----- Insert Table 7 about here -----

Control group versus outside group: assessing the effectiveness of funding alone

In Panel B of Table 7, we compare health centers in the control group versus health centers in the outside

¹⁹ Formally, the Mahalanobis distance δ between treated health center i and candidate health center j is given by $\delta = [(\mathbf{X}_i - \mathbf{X}_j)' \Sigma^{-1} (\mathbf{X}_i - \mathbf{X}_j)]^{1/2}$, where \mathbf{X} is the vector of matching characteristics and Σ the covariance matrix.

group. Since the control group receives PDSS funding, and the outside group does not, this comparison allows us to examine how funding as standalone (i.e., without governance improvements) affects health center outcomes.

As is shown, we find that funding as standalone leads to increases in the scale of the health center's operations—the number of employees increases by 13.0% (p -value = 0.001), the number of primary healthcare services by 52.4% (p -value = 0.015), the number of maternal and childhood healthcare services by 57.4% (p -value = 0.042), and the number of births by 38.6% (p -value = 0.088)—but does not bring about significant improvements in operating efficiency (column (1)) nor in the quality of healthcare services (columns (9)-(11)).

Taken together, the results from Tables 3 and 7 suggest that i) governance alone increases operating efficiency and social performance, but not the scale of the health centers (Table 3); ii) funding alone increases the scale of the health centers, but does not improve operating efficiency nor social performance (Panel B of Table 7); and iii) the combination of funding and governance leads to improvements in both the scale of the health centers, as well in health centers' operating efficiency and social performance (Panel A of Table 7). As such, these results suggest that funding is not a substitute for governance. Instead, they complement each other in improving the health centers' social impact in terms of both scale and performance.²⁰

7. Discussion and conclusion

How can non-profit organizations improve their governance in order to maximize their intended social and environmental impact? This question is of foremost importance not only for the non-profits themselves (as well as their donors and impact investors), but also for achieving the United Nations' SDGs. In short,

²⁰ In Table A9, we consider the same set of moderators as in Table 6, but with respect to the regressions of Table 7. In Panel A, we continue to find that i) density moderates the impact of the treatment on operating efficiency and infant mortality, ii) the increase in operating efficiency is higher for health centers with lower ex ante operating efficiency, and iii) the decrease in infant mortality is more pronounced for health centers with higher ex ante mortality. In addition, in both Panels A and B, we find that smaller health centers are more likely to increase their scale following the provision of PDSS funding.

improving the governance of the non-profit sector is important from an economic, environmental, and social perspective.

In this study, we shed light on this question by exploiting a randomized governance program conducted in the DRC's healthcare sector. The program was administered in about 1,000 non-profit health centers that were randomly assigned into a treatment and control group. While health centers in both groups received financial subsidies from the program, only those in the treatment group were subject to the “governance treatment” consisting of social performance-based incentives combined with A&F. As such, this randomized program provides an ideal setup to study how the adoption of governance mechanisms affects health centers' outcomes, holding everything else (including financial subsidies) constant.

We find that the governance treatment led to i) a significant increase in the number of services performed per employee, and ii) a significant reduction in the share of stillbirths and neonatal deaths. These findings indicate that governance plays an important role in improving non-profits' operating efficiency and social performance, respectively.

We further explore the underlying mechanism and document that the effect of the governance treatment is stronger in areas with a lower density of health centers (that is, areas where it is harder to adopt best practices from other clinics) and among health centers with lower (ex ante) operating efficiency and social performance, respectively, which points toward the importance of learning in explaining the treatment effect—that is, the health center staff learn how to improve their operations from the auditors' feedback, while being incentivized to incorporate and act on this feedback going forward.

Finally, we examine whether financial subsidies could potentially serve as a substitute for governance. We find that this is unlikely to be the case. Specifically, we find that health centers that only receive financial subsidies increase their scale, but do not show improvements in operating efficiency nor in the quality of their services. In contrast, health centers that receive both financial subsidies and the governance treatment improve both their scale as well as their operating efficiency and the quality of their services. This suggests that financial subsidies and governance operate as complements toward the objective of increasing the non-profit's overall social impact.

Our study contributes to several strands of the literature. First, by exploring the role of governance (in the form of social performance-based incentives and A&F) in enhancing the social impact of non-profits operating in a resource-constrained environment, this study relates to the large literature that has studied the role of governance and management practices in the context of for-profit organizations (e.g., Aguilera et al. 2016, Blader, Gartenberg, and Prat 2020, Bloom and Van Reenen 2007, Bloom et al. 2013, Bryan, Tilcsik, and Zhu 2017, Cai and Szeidl 2018, Chatterji et al. 2019, De Mel, McKenzie, and Woodruff 2014, Dimitriadis and Koning 2020, McKenzie 2021, McKenzie and Woodruff 2014, 2017). Since the non-profit and for-profit contexts are fundamentally different, it is a priori unclear i) which governance and management practices are available and effective for non-profits in a resource-constrained setting, as well as ii) when and where those practices matter more or less. Our study sheds light on these questions and highlights the effectiveness of a bundle of governance mechanisms that consists of social performance-based incentives combined with A&F. Moreover, we show that, although financial subsidies cannot substitute for governance, they can enhance the effectiveness of the governance bundle.

In addition, our study contributes to the literature that examines the social performance of i) for-profit organizations (e.g., Eccles, Ioannou, and Serafeim 2014, Flammer 2015, Flammer, Hong, and Minor 2019, King and Lenox 2001); ii) for-profit organizations in collaboration with non-profits and non-governmental organizations (e.g., Ballesteros and Gatignon 2019, Cabral et al. 2019, Chatain and Plaksenkova 2019, Durand and Huysentruyt 2021, Rousseau, Berrone, and Gelabert 2019); iii) hybrid organizations (e.g., Battilana and Dorado 2010, Cobb, Wry, and Zhao 2016, Jay 2013, Pache and Santos 2013, Quélin, Kivleniece, and Lazzarini 2017); iv) non-profit organizations in the healthcare sector (e.g., Kellogg 2009); as well as v) non-profit organizations in developing countries (e.g., Mair, Marti, and Ventresca 2012). Our study complements this vibrant line of work by exploring how the governance of non-profits (along with the interplay between governance and financial subsidies) can contribute to their social impact in a resource-constrained environment and the attainment of the United Nations' SDGs. In this regard, our study also adds to the growing literature that examines how management research can help

understand and address grand societal challenges (e.g., Berrone et al. 2016, George et al. 2016a, Vakili and McGahan 2016).

Moreover, our study examines an underexplored institutional context: Africa, and more specifically the DRC in Sub-Saharan Africa. While most of the insights from the governance literature have been obtained by studying organizations in developed and emerging countries, much less is known about (for-profit and non-profit) organizations in developing countries, and especially Africa (George et al. 2016b).²¹ Yet, understanding how to improve the governance of these organizations—and the implications for operating efficiency and social performance—is crucial in order to promote their development and the attainment of the United Nations’ SDGs.

Our study calls for future research. First, our findings are specific to the healthcare sector in a low-income country (namely, the DRC). In this regard, a fruitful avenue for future research is to examine whether our findings have external validity across sectors and countries, including higher-income countries. Arguably, the challenges faced by non-profit organizations in higher-income countries are similar but likely less severe than in low-income countries. As such, the effectiveness of the governance bundle considered in this study might differ. Future research may find it worthwhile to explore and characterize these differences. Second, a limitation of the PDSS program is that the “governance treatment” is administered through a bundle of governance mechanisms (i.e., social performance-based incentives combined with A&F), and hence we cannot separate between them. Accordingly, another exciting avenue for future research would be to “un-bundle” the governance bundle and examine how individual governance practices contribute to non-profits’ social impact. Third, and relatedly, future research could explore the effectiveness of other types of governance and management practices available to non-profit organizations.

Lastly—and perhaps most importantly—our findings have important implications for practice. Every year, large amounts of funding are invested in non-profit organizations pursuing social and

²¹ Similarly, the operations research literature in healthcare—which studies how to optimize hospitals’ operating processes—has focused on large-scale hospitals in the U.S. and other high-income countries (Berry Jaeker and Tucker 2016, Roth, Tucker, and Venkataraman 2019, Song et al. 2018).

environmental causes and aiming to achieve the seventeen SDGs of the United Nations (e.g., ending poverty, reducing hunger, promoting healthy lives and well-being, reducing inequalities, addressing climate change, protecting life on land and below water). The insights of this study help inform non-profit organizations, as well as their donors and impact investors, about the governance mechanisms that are available and effective in achieving the non-profits' objectives and maximizing the social impact of the funds invested.

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Figure 1. Examples of health centers

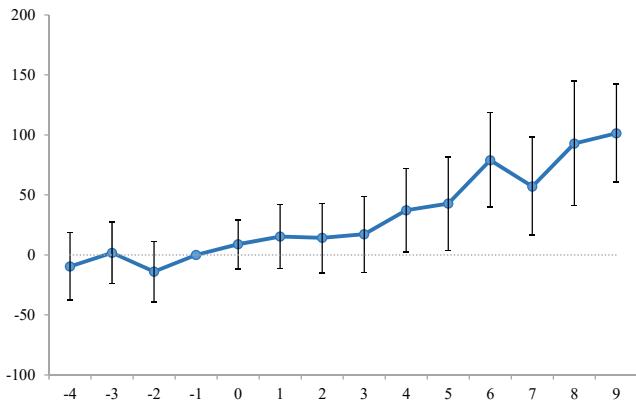


Source: Bluesquare.

Figure 2. Dynamics

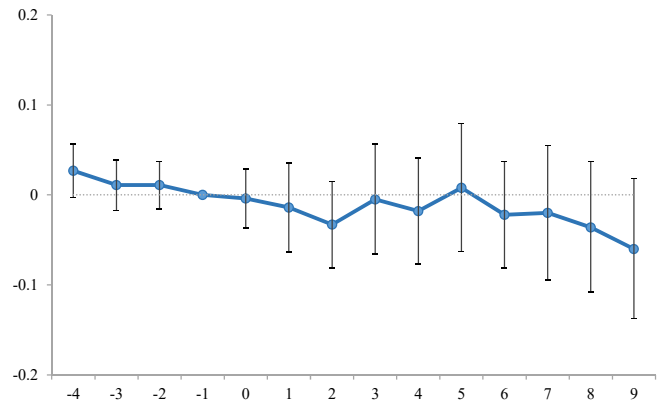
A. Primary healthcare services per employee

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



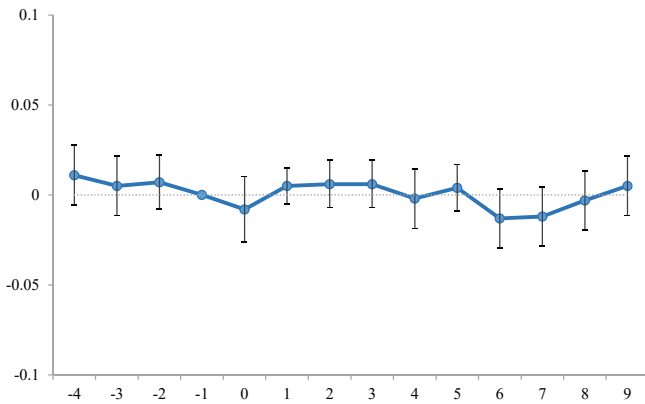
B. Employees

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



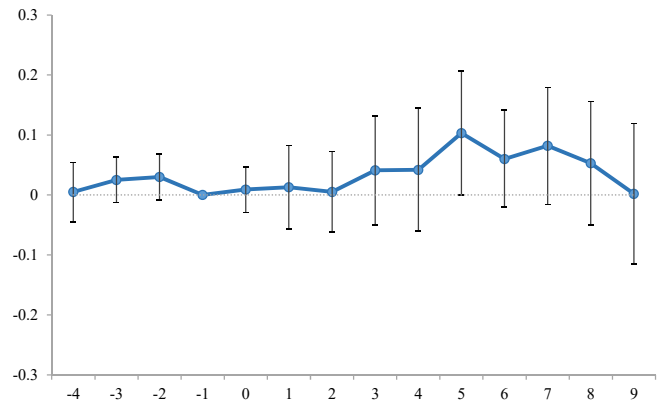
C. Doctors

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



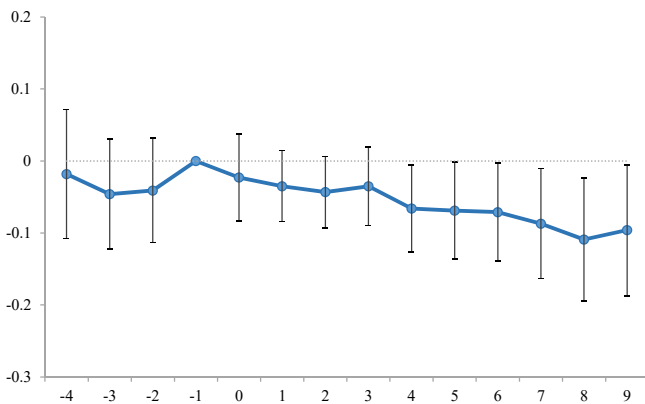
D. Nurses

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



E. Administrative employees

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



F. Primary healthcare services

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI

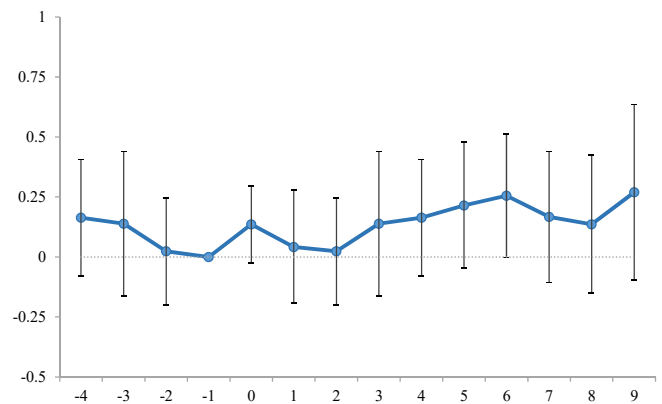
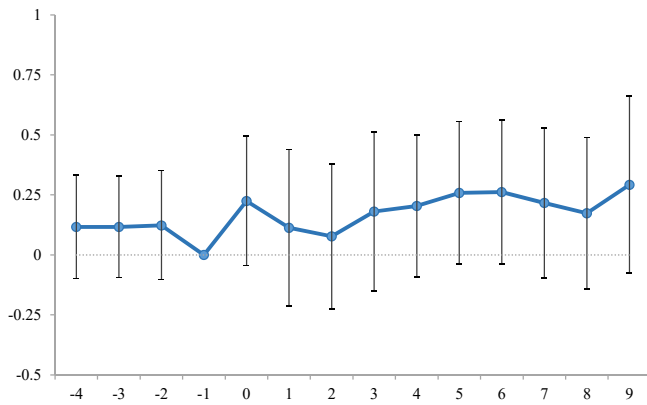


Figure 2 (continued)

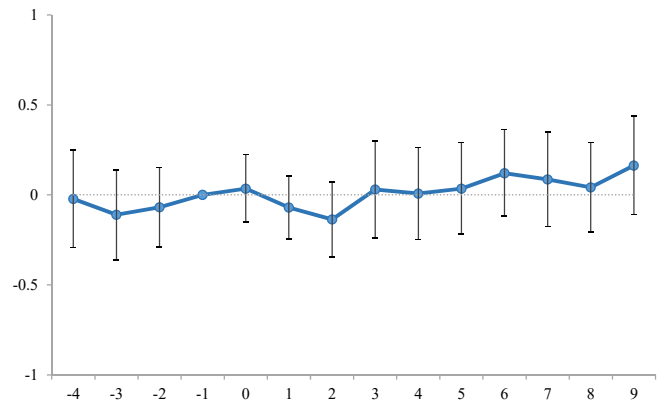
G. Maternal and childhood healthcare services

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



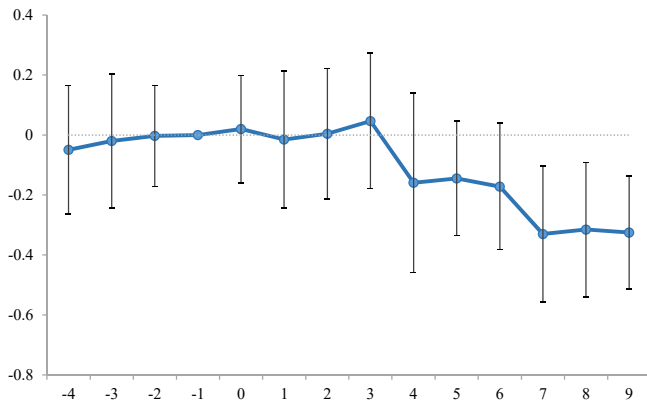
H. Births

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



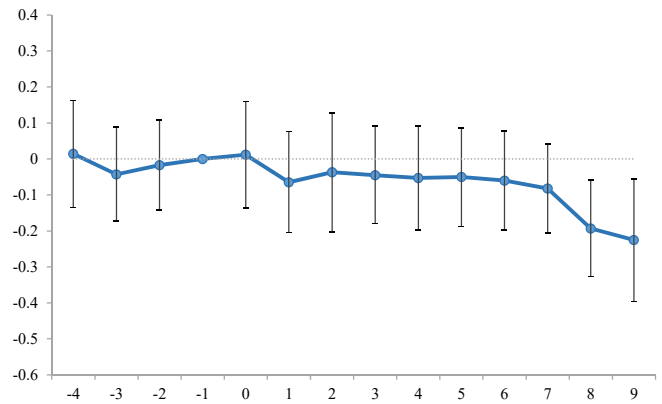
I. Share of stillbirths

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



J. Share of neonatal deaths

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI



K. Share of live births

Treatment vs. control, relative to Q4 2016 ($t = -1$), with 90% CI

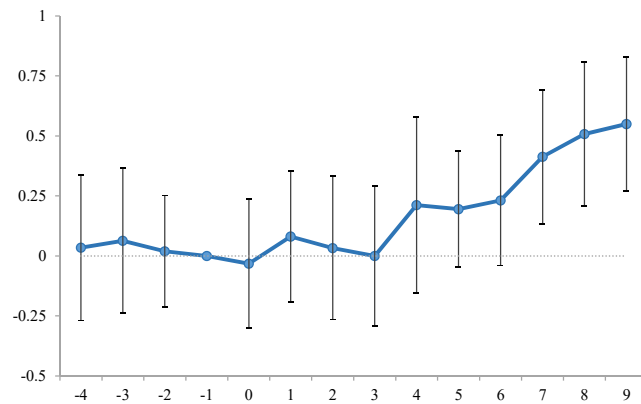


Table 1. Summary statistics

	Obs.	Mean	Median	Std. Dev.
Panel A. Health centers statistics				
Primary healthcare services per employee	999	235.76	184.88	214.01
Employees	999	7.13	6	6.30
Doctors	999	0.06	0	0.31
Nurses	999	3.55	3	3.70
Administrative	999	3.53	3	3.37
Primary healthcare services	999	1,611	1,344	1,195
Maternal and childhood healthcare services	999	957	705	940
Births	999	53.80	45	39.46
Stillbirths (in %)	999	0.70	0	1.51
Neonatal deaths (in %)	999	0.17	0	0.68
Live births (in %)	999	99.12	100	1.76
Panel B. Population statistics				
Population in center's health area	999	11,135	9,508	7,734
Population in center's health district	999	204,409	181,565	80,683

Table 2. Randomization tests

		Obs.	Mean	Median	Std. Dev.	<i>p</i> -value (diff. means)	<i>p</i> -value (KS test)
Panel A. Health centers statistics							
Primary healthcare services per employee	Treated	674	238.53	176.62	231.07	0.780	0.156
	Control	325	230.02	200.29	173.55		
Employees	Treated	674	7.43	6	7.21	0.196	0.140
	Control	325	6.50	6	3.67		
Doctors	Treated	674	0.06	0	0.33	0.457	0.885
	Control	325	0.04	0	0.25		
Nurses	Treated	674	3.79	3	4.26	0.135	0.675
	Control	325	3.05	3	2.03		
Administrative	Treated	674	3.58	3	3.66	0.710	0.453
	Control	325	3.42	3	2.68		
Primary healthcare services	Treated	674	1,660	1,361	1,273	0.351	0.213
	Control	325	1,508	1,338	1,009		
Maternal and childhood healthcare services	Treated	674	988	710	1,010	0.494	0.729
	Control	325	892	684	770		
Births	Treated	674	55.71	45	41.42	0.348	0.342
	Control	325	49.84	45	34.78		
Stillbirths (in %)	Treated	674	0.73	0	1.53	0.500	0.871
	Control	325	0.64	0	1.45		
Neonatal deaths (in %)	Treated	674	0.18	0	0.68	0.981	0.980
	Control	325	0.17	0	0.68		
Live births (in %)	Treated	674	99.09	100	1.772	0.584	0.863
	Control	325	99.18	100	1.749		
Panel B. Population statistics							
Population in center's health area	Treated	674	11,090	9,491	6,377	0.896	0.776
	Control	325	11,227	9,847	9,988		
Population in center's health district	Treated	674	201,829	181,565	73,925	0.716	0.458
	Control	325	209,760	177,275	93,047		

Notes. All variables are recorded in the quarter preceding the PDSS intervention (that is, Q4 2016). The penultimate column reports the *p*-value of the difference-in-means test comparing treated and control health centers. The last column reports the *p*-value of the Kolmogorov-Smirnov (KS) test of identical distributions comparing treated and control health centers. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table 3. The impact of the governance treatment on health center outcomes

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment	101.351*** (24.855)	-0.060 (0.047)	0.005 (0.010)	0.002 (0.071)	-0.096* (0.055)	0.270 (0.221)	0.292 (0.224)	0.163 (0.165)	-0.325*** (0.115)	-0.225** (0.103)	0.550*** (0.169)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.334	0.070	0.012	0.057	0.036	0.203	0.172	0.125	0.017	0.013	0.019
Observations	999	999	999	999	999	999	999	999	999	999	999

Notes. For each dependent variable we compute the change between $t - 1$ (the quarter preceding the intervention, Q4 2016) and $t + 9$ (nine quarters after the intervention, Q2 2019). When the dependent variable is a ratio—i.e., in columns (1) and (9)-(11)— Δy represents the difference in y from $t - 1$ to $t + 9$. When the dependent variable is a level—i.e., in columns (2)-(8)— $\% \Delta y$ represents the percentage change in y from $t - 1$ to $t + 9$. In column (1), the units are in number of primary healthcare services per employee; in columns (9)-(11), the units are in percentage points. Standard errors (reported in parentheses) are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table 4. Dynamics

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment ($\Delta t-4, t-1$)	-9.666 (16.975)	0.027 (0.018)	0.011 (0.010)	0.005 (0.030)	-0.018 (0.054)	0.164 (0.147)	0.117 (0.131)	-0.022 (0.164)	-0.049 (0.130)	0.014 (0.090)	0.035 (0.184)
Treatment ($\Delta t-3, t-1$)	1.726 (15.338)	0.011 (0.017)	0.005 (0.010)	0.025 (0.023)	-0.046 (0.046)	0.139 (0.183)	0.117 (0.128)	-0.110 (0.152)	-0.020 (0.135)	-0.043 (0.079)	0.063 (0.183)
Treatment ($\Delta t-2, t-1$)	-13.905 (15.326)	0.011 (0.016)	0.007 (0.009)	0.030 (0.023)	-0.041 (0.044)	0.024 (0.135)	0.123 (0.138)	-0.068 (0.133)	-0.003 (0.102)	-0.017 (0.075)	0.019 (0.142)
Treatment ($\Delta t-1, t$)	8.833 (12.354)	-0.004 (0.020)	-0.008 (0.011)	0.009 (0.023)	-0.023 (0.037)	0.136 (0.098)	0.224 (0.164)	0.035 (0.114)	0.020 (0.109)	0.012 (0.090)	-0.032 (0.162)
Treatment ($\Delta t-1, t+1$)	15.267 (16.172)	-0.014 (0.030)	0.005 (0.006)	0.013 (0.042)	-0.035 (0.030)	0.042 (0.143)	0.113 (0.197)	-0.070 (0.106)	-0.015 (0.139)	-0.065 (0.085)	0.081 (0.166)
Treatment ($\Delta t-1, t+2$)	14.142 (17.572)	-0.033 (0.029)	0.006 (0.008)	0.005 (0.041)	-0.043 (0.030)	0.024 (0.135)	0.077 (0.183)	-0.136 (0.126)	0.004 (0.131)	-0.037 (0.100)	0.033 (0.181)
Treatment ($\Delta t-1, t+3$)	17.192 (19.290)	-0.005 (0.037)	0.006 (0.008)	0.041 (0.055)	-0.035 (0.033)	0.139 (0.183)	0.181 (0.200)	0.030 (0.163)	0.046 (0.137)	-0.045 (0.082)	-0.000 (0.176)
Treatment ($\Delta t-1, t+4$)	37.163* (20.933)	-0.018 (0.036)	-0.002 (0.010)	0.042 (0.062)	-0.066* (0.037)	0.164 (0.147)	0.204 (0.180)	0.007 (0.154)	-0.159 (0.182)	-0.053 (0.088)	0.212 (0.223)
Treatment ($\Delta t-1, t+5$)	42.791* (23.566)	0.008 (0.043)	0.004 (0.008)	0.103 (0.063)	-0.069* (0.041)	0.215 (0.159)	0.258 (0.181)	0.035 (0.154)	-0.145 (0.116)	-0.050 (0.083)	0.195 (0.147)
Treatment ($\Delta t-1, t+6$)	78.975*** (23.841)	-0.022 (0.036)	-0.013 (0.010)	0.060 (0.049)	-0.071* (0.041)	0.256 (0.156)	0.262 (0.183)	0.121 (0.145)	-0.172 (0.128)	-0.060 (0.084)	0.232 (0.164)
Treatment ($\Delta t-1, t+7$)	57.070** (24.665)	-0.020 (0.045)	-0.012 (0.010)	0.082 (0.059)	-0.087* (0.046)	0.167 (0.166)	0.216 (0.189)	0.086 (0.159)	-0.330** (0.138)	-0.082 (0.075)	0.413** (0.169)
Treatment ($\Delta t-1, t+8$)	92.801*** (31.394)	-0.036 (0.044)	-0.003 (0.010)	0.053 (0.062)	-0.109** (0.052)	0.136 (0.174)	0.174 (0.192)	0.042 (0.152)	-0.315** (0.136)	-0.193** (0.081)	0.508*** (0.181)
Treatment ($\Delta t-1, t+9$)	101.351*** (24.855)	-0.060 (0.047)	0.005 (0.010)	0.002 (0.071)	-0.096* (0.055)	0.270 (0.221)	0.292 (0.224)	0.163 (0.165)	-0.325*** (0.115)	-0.225** (0.103)	0.550*** (0.169)

Notes. This table presents variants of the regressions in Table 3, in which different time intervals (from $t-4$ until $t+9$) are used to compute changes in the dependent variables relative to $t-1$. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table 5. Quality and quantity of healthcare services at other healthcare facilities in the same health district as the treated health centers

Panel A. Quality of healthcare services

	Hospitals in same health district as treated health centers			Control health centers in same health district as treated health centers			Outside health centers in same health district as treated health centers		
	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Mean	-0.060 (0.148)	-0.030 (0.085)	0.089 (0.192)	-0.032 (0.084)	-0.077 (0.092)	0.109 (0.151)	0.021 (0.045)	0.006 (0.019)	-0.027 (0.059)
Observations	121	121	121	72	72	72	1,192	1,192	1,192

Table 5 (continued)

Panel B. Volume of healthcare services

	Hospitals in same health district as treated health centers			Control health centers in same health district as treated health centers			Outside health centers in same health district as treated health centers		
	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Mean	0.100 (0.276)	0.067 (0.283)	0.146 (0.191)	-0.083 (0.121)	0.042 (0.272)	0.140 (0.231)	0.127 (0.188)	0.108 (0.169)	0.117 (0.079)
Observations	121	121	121	72	72	72	1,192	1,192	1,192

Notes. This table reports the mean of the dependent variable $\Delta y_{t-1,t+9}$ across healthcare facilities (hospitals in columns (1)-(3), control health centers in columns (4)-(6), and health centers outside the PDSS program in columns (7)-(9)) that are located within the same health district as the treated health centers. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table 6. Cross-sectional heterogeneity

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment	107.947*** (25.673)	-0.053 (0.042)	0.005 (0.009)	0.006 (0.068)	-0.091* (0.054)	0.296 (0.200)	0.320 (0.209)	0.159 (0.146)	-0.295** (0.115)	-0.195* (0.103)	0.489*** (0.172)
Treatment × Density of health centers	-26.072** (13.303)	0.016 (0.020)	-0.004 (0.005)	0.004 (0.041)	0.025 (0.020)	-0.031 (0.118)	-0.150 (0.148)	-0.041 (0.069)	0.073 (0.067)	0.244** (0.104)	-0.317** (0.142)
Treatment × Infant mortality	6.156 (5.758)	-0.004 (0.009)	-0.002 (0.005)	-0.001 (0.016)	-0.001 (0.011)	0.037 (0.029)	0.014 (0.027)	-0.043 (0.029)	-0.126 (0.093)	-0.103** (0.050)	0.229* (0.137)
Treatment × Size	-16.859 (15.777)	-0.049 (0.052)	0.015 (0.013)	-0.032 (0.073)	-0.056 (0.045)	0.034 (0.183)	-0.002 (0.197)	-0.008 (0.095)	0.039 (0.106)	-0.072 (0.103)	0.032 (0.174)
Treatment × Operating efficiency	-31.100* (17.931)	0.005 (0.025)	0.003 (0.008)	-0.063 (0.049)	0.023 (0.029)	-0.159 (0.112)	-0.197 (0.124)	0.038 (0.064)	-0.003 (0.060)	0.041 (0.039)	-0.038 (0.078)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.369	0.193	0.016	0.155	0.074	0.384	0.334	0.221	0.024	0.056	0.040
Observations	999	999	999	999	999	999	999	999	999	999	999

Notes. This table presents variants of the regressions in Table 3, interacting *treatment* with various characteristics measured prior to the PDSS intervention. For ease of interpretation, all interacted characteristics are standardized. All regressions also include the interacted variables as standalone controls. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table 7. Funding vs. governance

Panel A. Treatment group vs. outside group

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment vs. outside	86.233*** (18.102)	0.137*** (0.041)	-0.010 (0.010)	0.224*** (0.063)	0.031 (0.038)	0.897*** (0.264)	0.924*** (0.260)	0.691*** (0.149)	-0.247* (0.105)	-0.215** (0.084)	0.462** (0.187)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.315	0.130	0.025	0.109	0.049	0.164	0.137	0.160	0.019	0.024	0.022
Observations	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348

Table 7 (continued)

Panel B. Control group vs. outside group

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Control vs. outside	19.831 (23.906)	0.130*** (0.038)	-0.001 (0.009)	0.162** (0.064)	0.085* (0.047)	0.524** (0.211)	0.574** (0.278)	0.386* (0.224)	0.073 (0.174)	0.176 (0.160)	-0.250 (0.273)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.210	0.125	0.019	0.095	0.078	0.373	0.252	0.210	0.089	0.020	0.078
Observations	650	650	650	650	650	650	650	650	650	650	650

Notes. This table presents variants of the regressions in Table 3, except that the treatment group is compared to the outside group in Panel A, and the control group is compared to the outside group in Panel B. Health centers from the outside group are matched to health centers in the treatment and control groups, respectively, using the nearest-neighbor matching described in Section 6. Standard errors (reported in parentheses) are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Appendix

Appendix A. Structure of the DRC's health system

This appendix provides a brief description of the structure of the DRC's health system.²² The health system of the DRC is organized in three levels (central, provincial, and operational):

- At the *central level* (also referred to as “national level”), the Ministry of Health develops health policies and health standards. It plays a normative role, regulating and supporting the provincial health divisions.
- At the *provincial level* (also referred to as “intermediate level”), each province has a provincial ministry of health with its own provincial health division. Provincial health divisions are mainly responsible for the technical supervision and logistic support of health districts (at the operational level). They also manage the provincial hospital as well as other provincial health structures.
- At the *operational level*, the DRC is divided into 516 health districts (also called “health zones”) that cover a population of roughly 100,000 to 200,000. Each health district is administered by a health district management team (“*équipe cadre de la zone*”) that manages a network of health centers and the district hospital. Health districts are further divided into 8,504 health areas. Each health area contains approximately one health center.

Appendix B. Robustness

In this appendix, we describe several robustness checks that are variants of the baseline specifications presented in Table 3.

Specifications without province fixed effects. In Table A3, we re-estimate our baseline regressions without including province fixed effects, that is, without enforcing within-province comparisons of treated vs. control health centers. Due to the randomization, our point estimates should not be affected by the

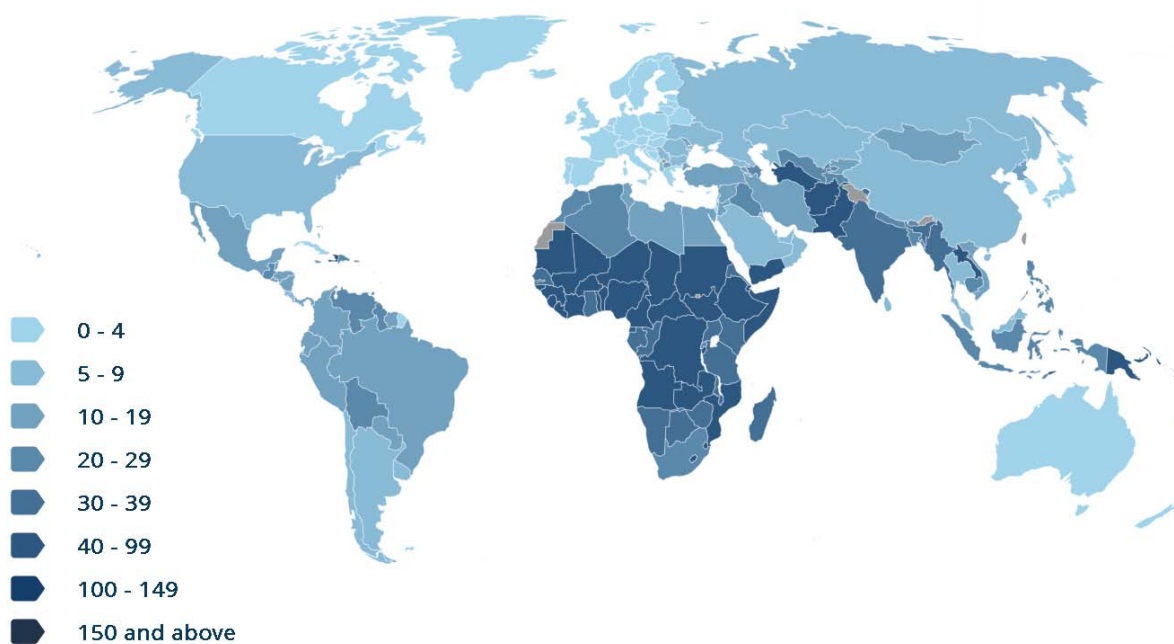
²² For a more detailed description, see WHO (2015).

inclusion of fixed effects. As can be seen, we indeed find that our results are very similar without including province fixed effects.

Specifications without clustered standard errors. In Table A4, we re-estimate our baseline regressions without clustering standard errors. As is shown, the statistical significance of our estimates is stronger without clustering.

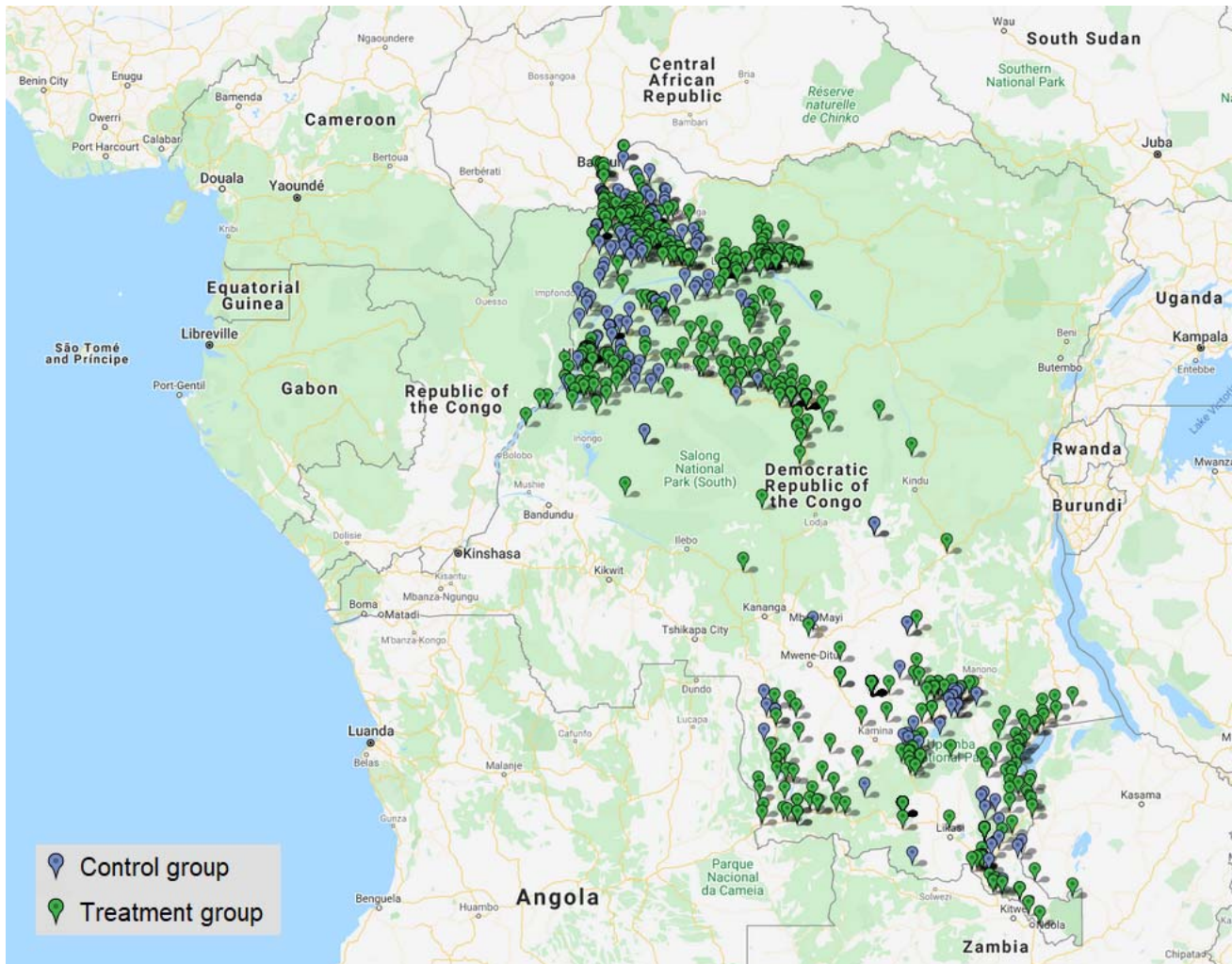
Weighted least squares. In Table A5, we re-estimate the regressions from columns (9)-(11) of Table 3—that is, the regressions pertaining to the number of stillbirths, the number of neonatal deaths, and the number of live births, respectively—using weighted least squares (WLS), weighting observations by the number of births. Doing so accounts for the fact that ratios of births are more accurately measured when the number of births is higher. As can be seen, we find that our results are slightly stronger when estimated by WLS.

Figure A1. Infant mortality rate



Notes. Infant mortality rate is measured by the number of deaths of infants under one-year old per 1,000 births alive. Source: United Nations Inter-Agency Group for Child Mortality Estimation (<https://childmortality.org/data>).

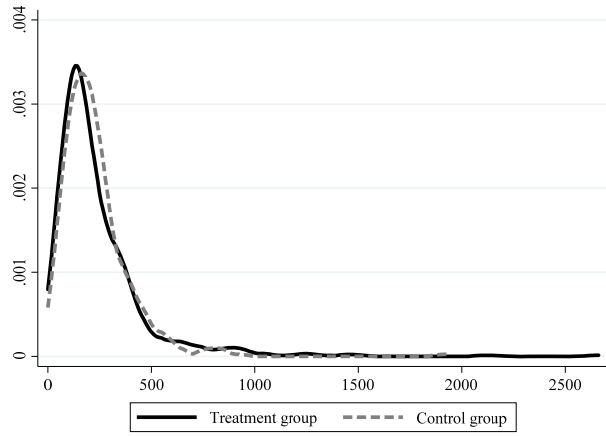
Figure A2. Location of treated and control health centers



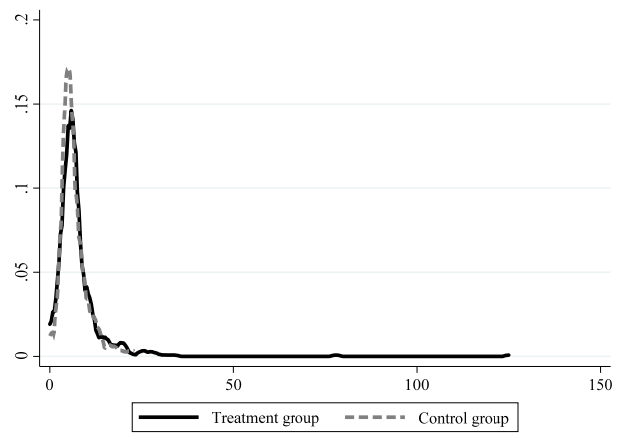
Notes. This figure plots the location of the health centers in the control group (blue markers) and treatment group (green markers), for the health centers in our sample that have non-missing geo-codes (73% of the sample). The map was generated using the online mapping software Maptive.

Figure A3. Pre-treatment distributions

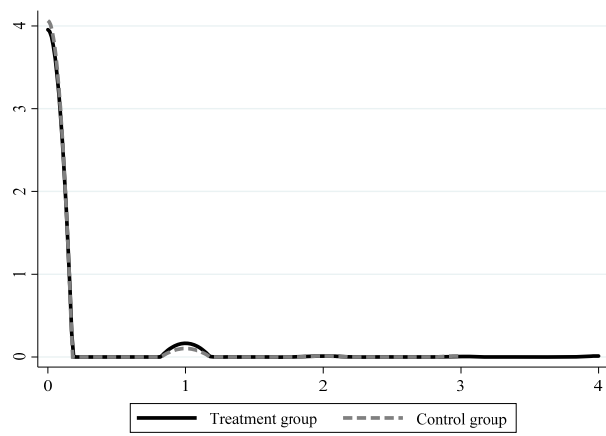
A. Primary healthcare services per employee at $t - 1$



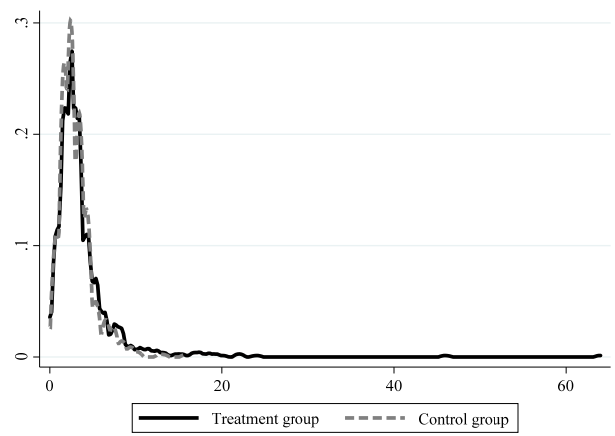
B. Employee at $t - 1$



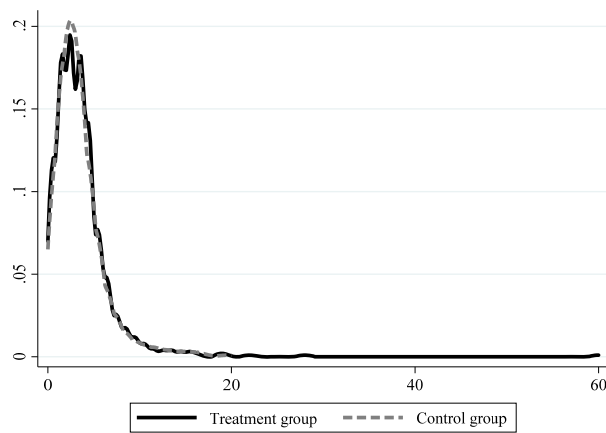
C. Doctors at $t - 1$



D. Nurses at $t - 1$



E. Administrative employees at $t - 1$



F. Primary healthcare services at $t - 1$

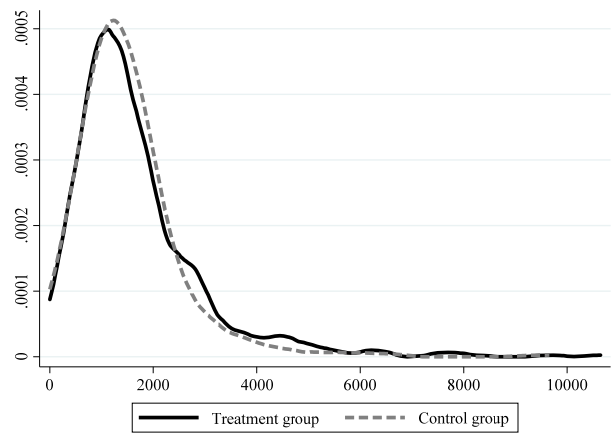
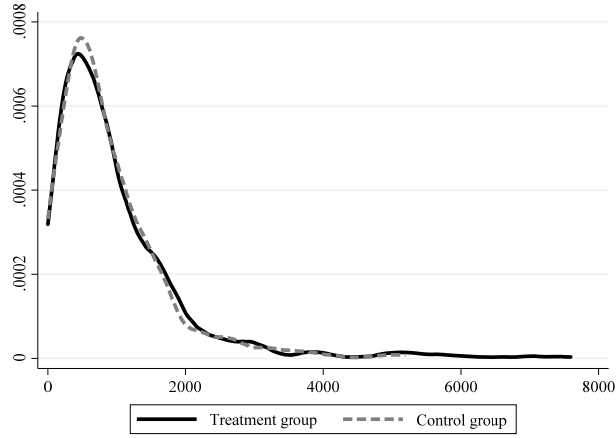
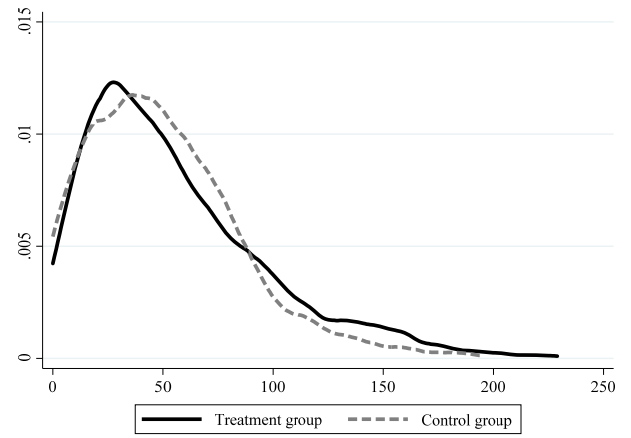


Figure A3 (continued)

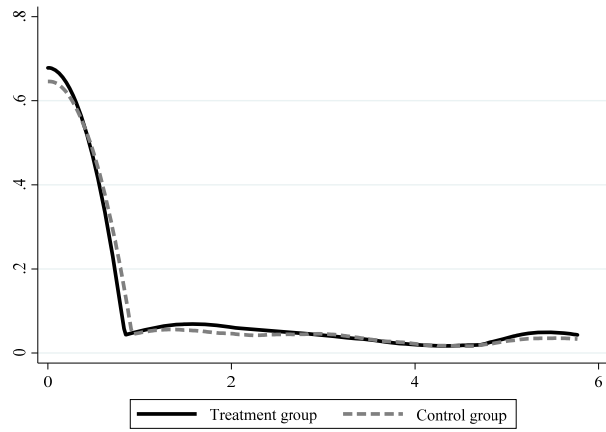
G. Maternal and childhood healthcare services at $t - 1$



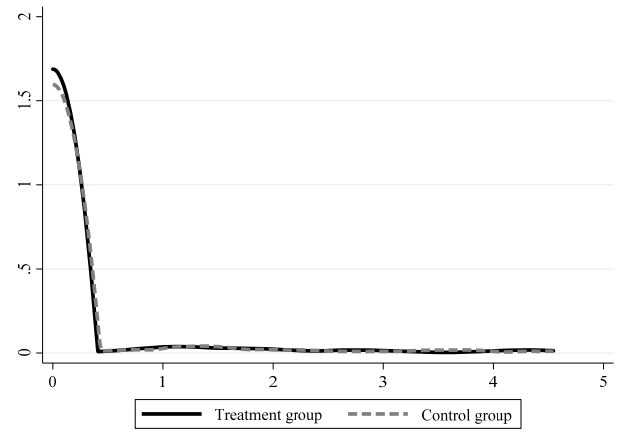
H. Births at $t - 1$



I. Share of stillbirths at $t - 1$



J. Share of neonatal deaths at $t - 1$



K. Share of live births at $t - 1$

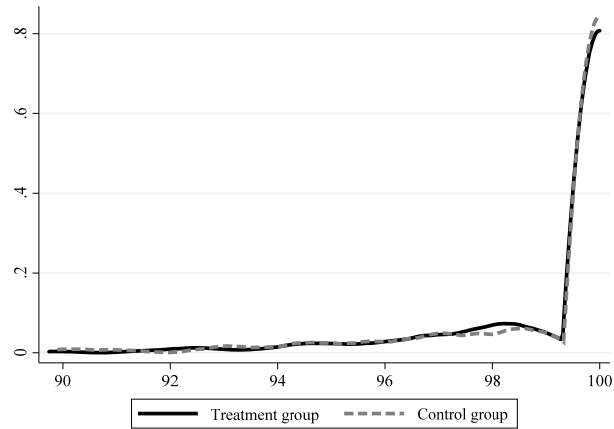
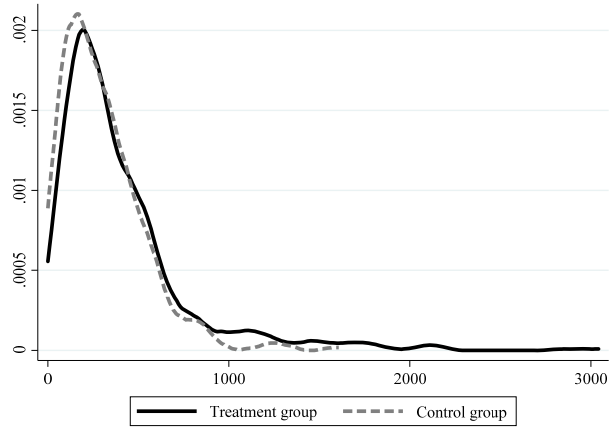
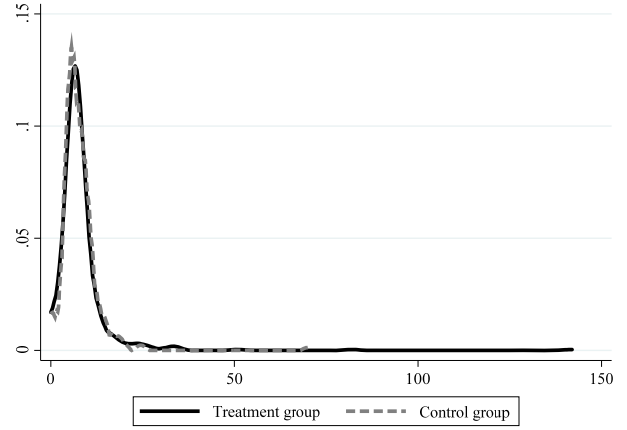


Figure A4. Post-treatment distributions

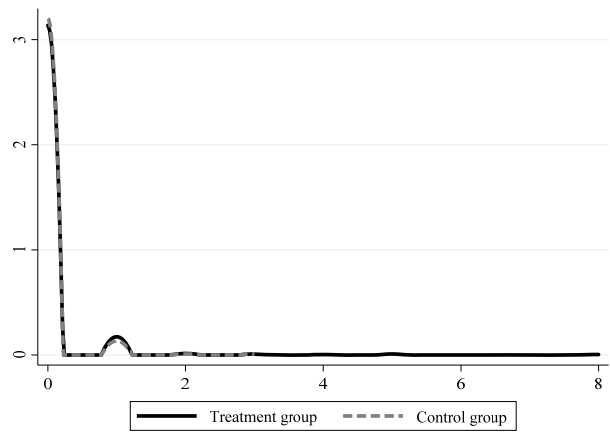
A. Primary healthcare services per employee at $t + 9$



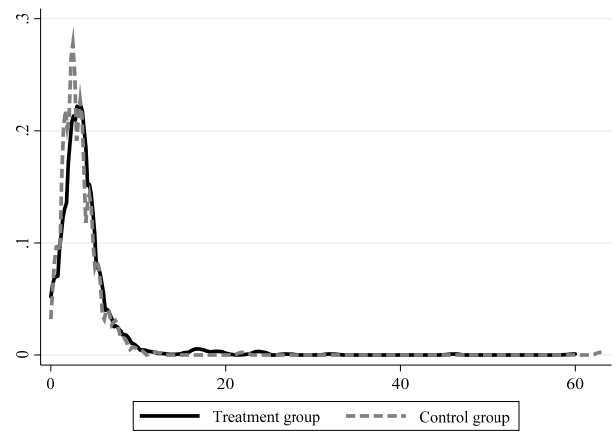
B. Employee at $t + 9$



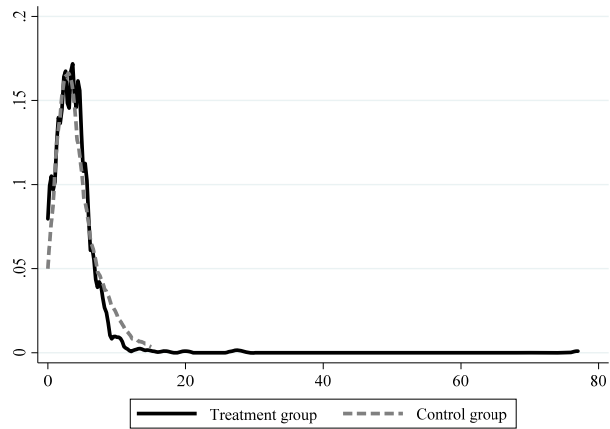
C. Doctors at $t + 9$



D. Nurses at $t + 9$



E. Administrative employees at $t + 9$



F. Primary healthcare services at $t + 9$

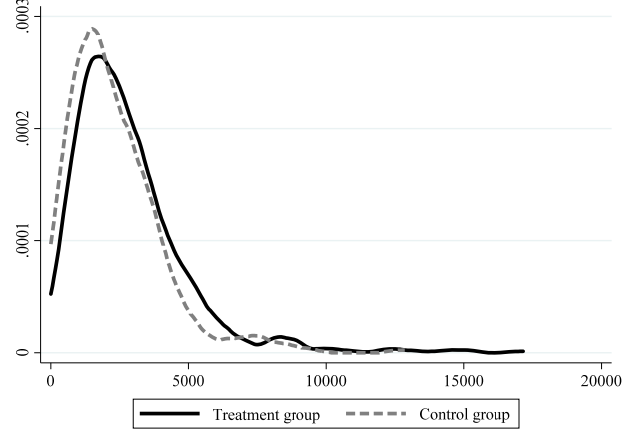
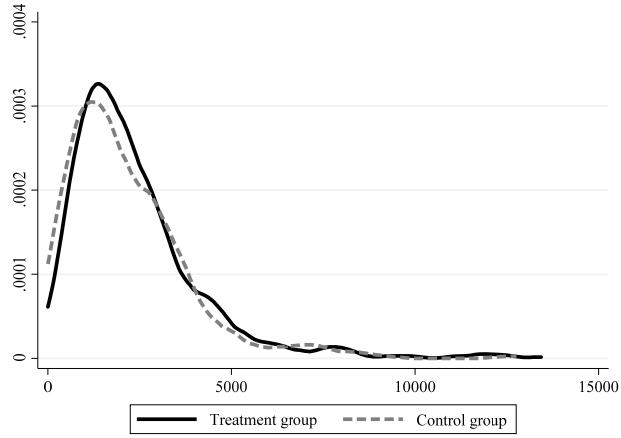
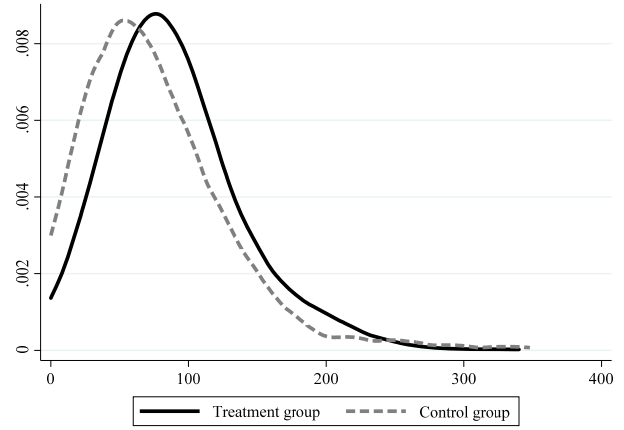


Figure A4 (continued)

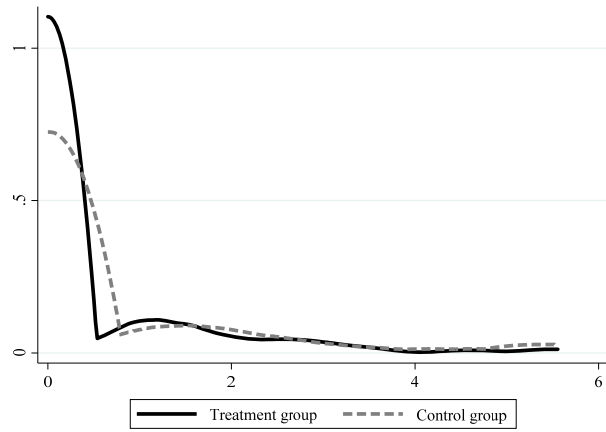
G. Maternal and childhood healthcare services at $t + 9$



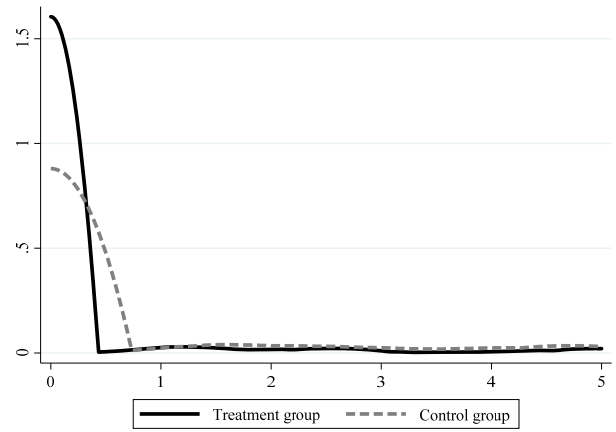
H. Births at $t + 9$



I. Share of stillbirths at $t + 9$



J. Share of neonatal deaths at $t + 9$



K. Share of live births at $t + 9$

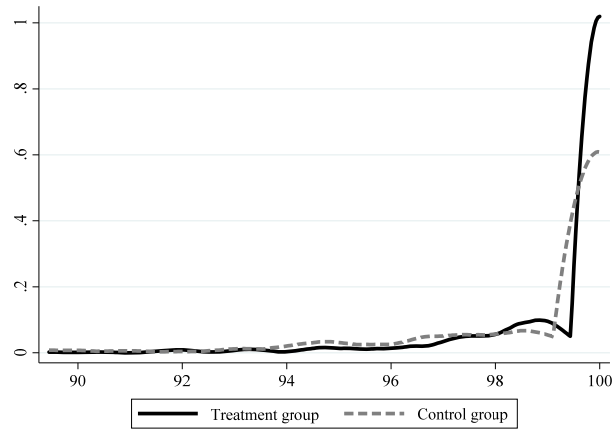


Table A1. Output indicators

1	Number of new curative care visits
2	Number of new curative care visits (vulnerable groups)
3	Number of minor surgeries
4	Number of severe cases referred to hospitals
5	Number of children immunized
6	Number of women receiving at least two doses of tetanus toxoid vaccine during prenatal care
7	Number of women receiving three doses of sulfadoxine-pyrimethamine during prenatal care
8	Number of first prenatal care visits
9	Number of fourth prenatal care visits
10	Number of second and third postnatal care visits
11	Number of assisted deliveries
12	Number of first-time family planning visits pertaining to short-acting contraceptive methods (injectable or oral)
13	Number of first-time family planning visits pertaining to long-term contraceptive methods (IUD or implants)
14	Number of growth-monitoring visits for children aged between 6 and 23 months (preventive care)
15	Number of growth-monitoring visits for children aged between 24 and 59 months (preventive care)
16	Number of home visits
17	Number of voluntary HIV counseling and testing visits
18	Number of HIV+ pregnant women receiving antiretroviral therapy
19	Number of HIV+ newborns receiving antiretroviral therapy
20	Number of HIV+ patients receiving antiretroviral therapy
21	Number of patients tested for pulmonary tuberculosis
22	Number of patients cured from pulmonary tuberculosis

Source: PDSS (2016, pp. 60-61).

Table A2. Weights for the quality score

	Maximum quality score	Weight
1 Overall organization	31	7.8%
2 Management plan	9	2.3%
3 Finance	15	3.8%
4 Indigent committee	20	5.0%
5 Hygiene and sterilization	31	7.8%
6 External consultations	128	32.0%
7 Family planning	32	8.0%
8 Laboratory	17	4.3%
9 Screening services	6	1.5%
10 Drugs and consumables	25	6.3%
11 Tracer drugs	17	4.3%
12 Maternity care	24	6.0%
13 Vaccination	20	5.0%
14 Pre-natal care	12	3.0%
15 HIV and tuberculosis care	13	3.3%
Total	400	100.0%

Source: PDSS (2016, p. 92).

Table A3. Robustness—specifications without fixed effects

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment	75.594** (34.444)	-0.037 (0.046)	0.007 (0.009)	-0.001 (0.066)	-0.063 (0.048)	0.355 (0.256)	0.267 (0.261)	0.225 (0.174)	-0.243** (0.122)	-0.224** (0.087)	0.467*** (0.163)
Province fixed effects	No	No	No	No	No	No	No	No	No	No	No
R-squared	0.027	0.002	0.001	0.000	0.005	0.009	0.005	0.010	0.005	0.009	0.011
Observations	999	999	999	999	999	999	999	999	999	999	999

Notes. This table reports variants of the regressions in Table 3, but without including province fixed effects. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table A4. Robustness—specifications without clustered standard errors

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment	101.351*** (12.182)	-0.060** (0.028)	0.005 (0.009)	0.002 (0.042)	-0.096*** (0.030)	0.270** (0.113)	0.292*** (0.112)	0.163** (0.072)	-0.325*** (0.117)	-0.225*** (0.085)	0.550*** (0.158)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.334	0.070	0.012	0.057	0.036	0.203	0.172	0.125	0.017	0.013	0.019
Observations	999	999	999	999	999	999	999	999	999	999	999

Notes. This table reports variants of the regressions in Table 3, but without clustering standard errors. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table A5. Robustness—birth-weighted regressions

	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)
Treatment	-0.358*** (0.118)	-0.253** (0.109)	0.611*** (0.179)
Province fixed effects	Yes	Yes	Yes
R-squared	0.020	0.014	0.022
Observations	999	999	999

Notes. This table provides variants of the regressions in columns (9)-(11) of Table 3, but estimating the regressions by weighted least squares (WLS), weighting observations by the number of births. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table A6. Low-performing health centers

	Health center operating efficiency		Quality of healthcare services	
	Δ Primary healthcare services per employee	Δ Primary healthcare services per employee	Δ Share of live births	Δ Share of live births
	(1)	(2)	(3)	(4)
Treatment	93.130*** (27.195)	96.631*** (25.858)	0.374** (0.176)	0.443** (0.175)
Treatment \times Operating efficiency (bottom 25%)	32.977* (19.805)			
Treatment \times Operating efficiency (bottom 10%)		48.327* (28.082)		
Treatment \times Infant mortality (bottom 25%)			0.622** (0.305)	
Treatment \times Infant mortality (bottom 10%)				0.864* (0.505)
Province fixed effects	Yes	Yes	Yes	Yes
Controls	Yes	Yes	Yes	Yes
R-squared	0.337	0.337	0.028	0.029
Observations	999	999	999	999

Notes. This table presents variants of the regressions in columns (1) and (11) of Table 3, interacting *treatment* with two dummy variables that indicate whether operating efficiency and infant mortality, respectively, are in the bottom quartile (“bottom 25%”) and bottom decile (“bottom 10%”) of their distribution prior to the PDSS intervention. The regressions include the interacted variables as standalone controls. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table A7. Summary statistics for the outside group

	Obs.	Mean	Median	Std. Dev.
Panel A. Health centers statistics				
Primary healthcare services per employee	5,832	168.10	80.07	429.14
Employees	5,832	5.20	4	6.64
Doctors	5,832	0.23	0	0.86
Nurses	5,832	2.82	2	4.69
Administrative	5,832	2.15	1	2.59
Primary healthcare services	5,832	1,153	421	2,843
Maternal and childhood healthcare services	5,832	607	262	979
Births	5,832	26.77	13	41.39
Stillbirths (in %)	5,832	0.41	0	1.30
Neonatal deaths (in %)	5,832	0.37	0	1.95
Live births (in %)	5,832	99.22	100	2.55
Panel B. Population statistics				
Population in center's health area	5,832	11,918	9,495	8,700
Population in center's health district	5,832	215,158	194,315	94,966

Table A8. Covariate balance post matching

	Treatment vs. matched outside group		Control vs. matched outside group	
	Obs.	<i>p</i> -value (diff. in means)	Obs.	<i>p</i> -value (diff. in means)
	(1)	(2)	(3)	(4)
Primary healthcare services per employee	1,348	0.342	650	0.516
Employees	1,348	0.659	650	0.638
Doctors	1,348	0.402	650	0.485
Nurses	1,348	0.433	650	0.822
Administrative	1,348	0.950	650	0.512
Primary healthcare services	1,348	0.878	650	0.820
Maternal and childhood healthcare services	1,348	0.861	650	0.622
Births	1,348	0.414	650	0.649
Stillbirths (in %)	1,348	0.518	650	0.343
Neonatal deaths (in %)	1,348	0.420	650	0.790
Live births (in %)	1,348	0.822	650	0.375

Notes. All variables are recorded in the quarter preceding the PDSS intervention (that is, Q4 2016). The table reports the *p*-value of the difference-in-means test comparing treated health centers vs. matched outside health centers (columns (1) and (2)), and control health centers vs. matched outside health centers (columns (3) and (4)). Health centers from the outside group are matched to health centers in the treatment and control groups, respectively, using the nearest-neighbor matching described in Section 6. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.

Table A9. Funding vs. governance—cross-sectional heterogeneity

Panel A. Treatment group vs. outside group

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Treatment vs. outside	86.873*** (18.005)	0.165*** (0.034)	-0.005 (0.011)	0.278*** (0.060)	0.045 (0.036)	0.826*** (0.231)	0.817*** (0.187)	0.631*** (0.112)	-0.307*** (0.113)	-0.277*** (0.080)	0.584*** (0.213)
Treatment vs. outside × Density of health centers	-19.744** (7.891)	0.041 (0.039)	0.006 (0.010)	0.032 (0.051)	0.029 (0.020)	-0.144 (0.293)	-0.119 (0.304)	-0.118 (0.159)	0.198* (0.116)	0.176** (0.083)	-0.374** (0.143)
Treatment vs. outside × Infant mortality	-8.794 (11.338)	-0.003 (0.043)	0.012 (0.012)	0.011 (0.050)	-0.026 (0.038)	-0.095 (0.208)	0.046 (0.248)	-0.081 (0.113)	-0.115 (0.087)	-0.191* (0.097)	0.306** (0.136)
Treatment vs. outside × Size	-13.342 (10.393)	-0.127*** (0.038)	0.012 (0.018)	-0.137** (0.053)	-0.104** (0.049)	-0.466** (0.185)	-0.458* (0.250)	-0.281** (0.141)	-0.032 (0.134)	0.046 (0.087)	-0.013 (0.234)
Treatment vs. outside × Operating efficiency	-23.522* (13.898)	0.024 (0.024)	0.004 (0.007)	0.032 (0.040)	0.044 (0.028)	-0.162 (0.310)	-0.173 (0.322)	-0.129 (0.139)	-0.078 (0.064)	-0.044 (0.113)	0.121 (0.138)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.555	0.209	0.063	0.184	0.076	0.291	0.244	0.229	0.090	0.094	0.097
Observations	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348	1,348

Table A9 (continued)

Panel B. Control group vs. outside group

	Health center operating efficiency	Health center employees				Volume of healthcare services			Quality of healthcare services		
	Δ Primary healthcare services per employee	%Δ Emp.	%Δ Doctors	%Δ Nurses	%Δ Admin. employees	%Δ Primary healthcare services	%Δ Maternal and childhood healthcare services	%Δ Births	Δ Share of stillbirths	Δ Share of neonatal deaths	Δ Share of live births
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
Control vs. outside	17.769 (13.971)	0.138*** (0.033)	-0.001 (0.010)	0.183*** (0.066)	0.092** (0.044)	0.441** (0.183)	0.494** (0.246)	0.393** (0.181)	0.104 (0.178)	0.139 (0.129)	-0.243 (0.211)
Control vs. outside × Density of health centers	-13.892 (11.030)	-0.033 (0.041)	0.011 (0.010)	-0.051 (0.060)	-0.049 (0.032)	-0.039 (0.103)	-0.031 (0.145)	-0.156 (0.145)	-0.160 (0.155)	-0.130 (0.120)	0.291 (0.218)
Control vs. outside × Infant mortality	-4.375 (12.590)	-0.036 (0.026)	-0.014 (0.009)	-0.045 (0.040)	-0.012 (0.032)	-0.037 (0.136)	0.030 (0.273)	-0.033 (0.137)	-0.167 (0.154)	-0.137 (0.167)	0.303 (0.236)
Control vs. outside × Size	-0.597 (6.436)	-0.105** (0.047)	-0.014 (0.010)	-0.124** (0.058)	-0.091** (0.042)	-0.195* (0.112)	-0.240* (0.142)	-0.283** (0.116)	0.110 (0.139)	-0.019 (0.128)	-0.091 (0.234)
Control vs. outside × Operating efficiency	-21.640 (23.147)	-0.024 (0.054)	0.009 (0.006)	0.037 (0.075)	-0.063 (0.060)	-0.230 (0.180)	-0.239 (0.260)	0.172 (0.181)	-0.132 (0.133)	0.080 (0.127)	0.052 (0.109)
Province fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
R-squared	0.573	0.196	0.036	0.174	0.154	0.560	0.388	0.365	0.153	0.111	0.181
Observations	650	650	650	650	650	650	650	650	650	650	650

Notes. This table presents variants of the regressions in Table 7, interacting *treatment vs. outside* and *control vs. outside* with various characteristics measured prior to the PDSS intervention. For ease of interpretation, all interacted characteristics are standardized. All regressions also include the interacted variables as standalone controls. Standard errors are clustered at the health district level. *, **, and *** denotes significance at the 10%, 5%, and 1% level, respectively.