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for Global Education

Building resilient education systems: Experimental evidence across five countries

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February 2026



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The What Works Hub for Global Education is an international partnership, funded by the UK government's Foreign, Commonwealth & Development Office and the Gates Foundation, working out how to effectively implement education reforms at scale.

Please cite this as:

Angrist, N., Ainomugisha, M., Bathena, S.P., Bergman, P., Crossley, C., Cullen, C., Letsomo, T., Matsheng, M., Panti, R.M., Sabarwal, S. & Sullivan, T. 2025. Building resilient education systems: Experimental evidence across five countries. What Works Hub for Global Education Working Paper Series. 2026/002.

https://doi.org/10.35489/BSG-WhatWorksHubforGlobalEducation-WP_2026/002

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Building Resilient Education Systems: Experimental Evidence Across Five Countries

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Abstract

Education systems need to withstand frequent shocks, including disease and climate events, which close schools for over 222 million children. During these emergencies, alternative models are needed to deliver education. However, rigorous evaluation of effective educational approaches in these settings is challenging and rare, especially across multiple countries. We present results from five randomized trials in India, Kenya, Nepal, Philippines, and Uganda to evaluate the provision of education in emergency settings. These multi-site coordinated trials enable us to uniquely inform internal and external validity simultaneously. We also test multiple scalable models of remote targeted tutoring instruction, comparing government and NGO delivery. While SMS messages have mixed results, phone call tutorials show consistent effectiveness across diverse settings. We find large and robust effect sizes on learning, with average effects of 0.30-0.35 standard deviations. These effects are highly cost-effective, delivering up to four years of high-quality instruction per \$100 spent. Results also show similar effects when delivered by government teachers as NGO instructors and sustained improvements to teacher beliefs and practices. We further evaluate effects during a government scale-up with 25,000 teachers in India and an additional weather shock in the Philippines. Altogether, our results reveal it is possible to strengthen the resilience of education systems, enabling education provision amidst disruptions, and to deliver cost-effective learning gains across contexts and with governments.

Keywords: Human Capital, Education, Education in Emergencies, External Validity, Scale

*We are grateful to a coalition of partners who enabled this multi-country effort. Implementing & research partners include: Youth Impact, J-PAL, Learning Collider (supported by Schmidt Futures & Citadel), Oxford University, World Bank, Nepal's Ministry of Education, Science and Technology, Teach for Nepal, Street Child, Philippines Department of Education, IPA, Building Tomorrow, NewGlobe, Alokit, and Global School Leaders. Funders include UBS Optimus Foundation, Mulago Foundation, Douglas B. Marshall Foundation, Echidna Giving, Stavros Niarchos Foundation, Jacobs Foundation, JPAL IGI, Northwestern's 'Economics of Nonprofits' class, the Peter Cundill Foundation, and the World Bank. Cullen received funding from the UKRI GCRF & Oxford's OPEN Fellowship. We thank Natasha Ahuja, Amy Jung, and Rachel Zhou who provided excellent research assistance. Special thank you to implementation and research teams including Nassreena Baddiri, Mariel Bayangos, Kshitiz Basnet, John Lawrence Carandang, Clotilde de Maricourt, Pratik Ghimire, Faith Karanja, Manoj Karki, Karisha Cruz, Usha Limbu, Roger Masapol, Edward Munyaneza, Sunil Poudel, Karthika Radhakrishnan-Nair, Pratigya Regmi, Uttam Sharma, Swastika Shrestha, and Kishor Subedi. We thank conference participants from ASSA, BREAD, CSAE, NEUDC, PacDev for comments. We thank Michela Carlana, Dean Karlan, and Mike Callen for detailed comments. IRB approval was granted by Nepal's Health Research Council (776/2020), HML IRB (889OXFD21), and Columbia University IRB (21-384). The trials were registered in the AEA registry: AEARCTR-0009779. For the India scale-up effort, we thank the Karnataka government, J-PAL South Asia, Alokit, the Agency Fund, the What Works Hub for Global Education, the UK Foreign Commonwealth and Development Office (FCDO) as well as Sophie Ochmann, Rohan Katepallewar, Subash Prabakaran, Kavya Manjunath, Namitha Sadanand, Rijul Grover, Swathi Attavar, & Sharnic Djaker. We dedicate this paper in loving memory to Micheal Ainomugisha. His commitment to evidence and improving children's lives was unwavering.

I Introduction

More than 2 billion people live in countries affected by emergencies that frequently disrupt education. While frequent and disruptive, education emergencies have historically been understudied. We present a new database documenting just how frequent and disruptive such shocks can be.¹ Schools close for lengthy periods during these emergencies, and learning loss can be substantial (Andrabi, Daniels, and Das 2021; Lichand et al. 2022; Carlana, La Ferrara, and Lopez 2023). Education Cannot Wait, the United Nation’s Global Fund for education in emergencies, estimates that 222 million children experience regular schooling disruptions and are in active need of education in emergencies programs.² Moreover, school breaks, while not emergencies, are another regular school disruption, often resulting in substantial learning loss among low-resource households (Cooper et al. 1996; Slade et al. 2017). These shocks exacerbate a pre-existing learning crisis, with many students well behind grade-level expectations (Ganimian and Djaker 2022). Fewer than half of primary-age students in low- and middle-income countries are able to read a story or perform two-digit math operations (World Bank 2018).

Resilient education systems need to promote learning across settings and to withstand frequent disruptions. Ideally, interventions that promote learning in emergencies are low cost, simple to implement, and quick to deploy. They should also yield high take-up across different geographies and implementation models and should be targeted to children of diverse educational and cultural backgrounds. Understanding which approaches can be effective across these circumstances requires multi-context, multi-model studies. However, rigorous evaluation of approaches to deliver education in emergencies remains challenging and rare, especially across contexts and with governments.

In this paper, we evaluate a set of education in emergencies programs to promote learning during large-scale school disruptions, which affected over 1 billion children worldwide. We conduct five randomized controlled trials, including multiple delivery models, such as NGO teacher aides as well as government teachers, to test scalability by government systems. Our trials were conducted across five countries: India, Kenya, Nepal, Uganda, and the Philippines. In all settings, schooling was disrupted due to the COVID-19 pandemic, and several of the countries in our study experienced some of the longest school closures in the world. In addition, a typhoon occurred during the trials in the Philippines; we evaluate persistence of effects during this additional education emergency. We further evaluate a post-pandemic scale-up in India to assess effectiveness in an additional setting and at an order of magnitude larger scale.

We use mobile phones to provide various educational interventions to primary school children. Mobile phones provide a platform that can cheaply reach households at scale in low-resource con-

¹Figure 1 documents the extent of school disruption for a selected set of emergencies over the last two decades. Notable examples include large earthquakes in Pakistan in 2005 and in Haiti in 2010, elections in Nepal in 2017 and air pollution in 2021, Ebola in Liberia and Sierra Leone, floods in Bangladesh in 2007 and Monsoon rains in 2019, prolonged droughts in Kenya in 2017, and foot and mouth disease in 2004 in Cambodia, among others. Conflicts also have substantial negative effects on education. For example, between 2015 and 2019 alone, more than 11,000 attacks on schools were documented in at least 93 countries (GCPEA, Education under Attack 2020).

²In low-income countries, students lose on average 10 percent of the academic year due to climate shocks alone (Sabarwal et al. 2024).

texts (Aker and Mbiti 2020). While less than 15 percent of households have access to the internet in low-income countries, over 70 percent have access to mobile phones (Carvalho and Crawford 2020). Moreover, mobile phones enable teachers to reach students at home even when school is disrupted, providing a resilient and flexible modality to provide education during emergencies. Mobile phones are increasingly relied on to provide aid via mobile money during humanitarian crises (Callen et al. 2025), yet remain underutilized to provide education. One treatment included a set of SMS messages, such as numeracy content provided weekly, as well as nudges to engage in educational activities. A second treatment provided additional weekly 20 minute phone call tutorials for eight weeks. The educational pedagogy was as essential as the mobile phone platform. Phone calls covered foundational numeracy and aimed to target instruction to student learning levels via regular low-cost, high-frequency assessments. This builds on targeted instruction approaches in-person and using technology (Banerjee et al. 2007; Banerjee et al. 2017; Muralidharan, Singh, and Ganimian 2019; Duflo, Kiessel, and Lucas 2020; Ganimian et al. 2020).

A proof of concept in Botswana showed phone call tutorials were effective in promoting learning during initial pandemic school disruptions (Angrist, Bergman, and Matsheng 2022). However, questions remain on whether this approach can be scaled across contexts and when delivered by governments – a pervasive challenge for social programs (List 2022; Mobarak 2022). Indeed, while the initial proof of concept provides high internal validity, it does not address critical questions around external validity. To ensure scientifically robust findings, we need both, yet external validity often remains elusive. By conducting large randomized trials across five contexts as well as comparing models, such as NGO and government delivery, we substantially enhance inference on both internal and external validity, as well as explicitly test more scalable models, such as government teacher delivery. Moreover, we also include evidence from multiple types of school disruptions, including a typhoon, and quasi-experimental estimates of effectiveness at scale post-pandemic.

Results show consistently large and robust effect sizes of phone call tutorials on learning across contexts, with average effects across all five countries of 0.30-0.35 standard deviations. These results translate into large learning gains in absolute terms. In Uganda, for example, less than 20 percent of grade 4 students can divide at baseline, but by endline, nearly 50 percent can. These gains fully recover learning losses in math and enable substantial progress beyond status quo learning rates. On average, across all countries, we find a 65 percent increase in the share of students who learn division. The effects are largest for students whose caregivers have only a primary education (rather than secondary and beyond), suggesting that results are strongest when there are fewer alternative educational support systems at home. Additional results show positive effects of phone call tutorials on learning higher-order competencies, such as fractions. Since fractions were not directly taught during the intervention, this provides evidence that learning extended beyond familiarity with the content taught. This reveals dynamic complementarities, with the benefits of learning basic numeracy accruing to learning additional skills (Cunha and Heckman 2007).

We randomized whether the intervention was provided by NGO instructors or government teachers in the Philippines and Nepal. The average effect of phone call tutorials on learning when

delivered by NGOs is 0.26 standard deviations and 0.31 when delivered by government teachers. These results show similar and statistically indistinguishable effects, indicating that government actors can effectively deliver these types of education in emergency responses. We also find high engagement across sites ranging from 70 to 80 percent, revealing the accessibility and robustness of the approach even in disrupted and low-resource environments.

We also embedded a study randomly allocating a subset of government teachers to deliver phone call tutorials in Nepal. This experimental variation enables detection of the impact of delivering the program on teacher beliefs and practices – an effect that could spill over into the education system and persist beyond the intervention. Results shows teacher practices shift substantially; teachers are 9.3 percentage points more likely to target their feedback to students’ learning level. Teachers are also more likely to get parents involved in education. We find large effects on teacher perceptions that they were able to help students learn, as well as their desire to teach, with a 15.8 percentage point gain in wanting to be a teacher if they could make the choice again. These results suggest that delivering effective programs could potentially unlock a virtuous cycle within government education systems, in turn motivating teachers to want to teach and to improve their teaching practice. Of note, the teachers in this trial closely resemble the average primary school teacher, suggesting government delivery results are likely to translate to a broader set of teachers in the education system.³

In contrast to phone call tutorials, the effects of SMS messages alone are mixed. On average, we find a 0.08 standard deviation effect on learning. While these effects are positive and statistically significant when pooled across contexts, they are not consistently statistically significant by country. Average effects are driven by substantial impacts in Uganda, with a 0.20 standard deviation effect that is significant at the 99 percent level, as well as effects in the Philippines, with an effect of 0.09 that is significant at the 90 percent level; there is no effect in Kenya or Nepal. These results suggest that SMS messages can work in contexts with the largest need, such as Uganda and the Philippines, but not in all contexts. Live phone call instruction, on the other hand, appears to strike the balance of being intensive enough to deliver sustained impact across diverse contexts while remaining cheap and scalable.⁴

³Teachers in the study are likely representative of a broader set of teachers since all were government teachers, teachers were directly requested to participate through government protocols by the Ministry of Education, and 80 percent of teachers expressed interest in the program and were enrolled into the eligible pool for the study. Moreover, given in this trial we introduce further randomization—teachers randomly assigned to implementation—the set of implementers are representative of the overall eligible pool of teachers.

⁴We also contribute to the development of remote learning assessments. High-frequency remote assessment data enabled real-time targeting of instruction to student levels as well as evaluation of program effectiveness. Even prior to the pandemic, phone calls have been used for household surveys, such as the World Bank Living Measurement Study (LSMS) and UNICEF’s Multiple Indicator Cluster Survey (MICS). We conduct five validity checks on high-frequency, low cost learning assessments via phone. A first check compares in-person to phone-based assessment for the exact same set of students in Kenya. We find no statistically distinguishable difference between these two modes of assessments. An additional test included back-checks, with a random subset of students tested twice on the same competencies. We find a strong relationship as expected. We additionally randomize various problems of the same proficiency (e.g., four different questions to measure 2-digit addition with carryover). Results show no difference by question, showing accurate estimates of latent ability. Finally, we include a real-effort question to disentangle effects of the intervention on effort on the test, which has been shown to affect test scores during in-person exams (Gneezy

Beyond the core set of randomized trials that occurred during COVID-19 school disruptions, an additional education emergency took place during the course of the study: a devastating typhoon in the Philippines that destroyed 4,000 classrooms and disrupted learning for 2 million children (OCHA 2022). The typhoon resulted in a shock that further disrupted schooling and learning. Our initial randomization remained unbiased among groups affected by the typhoon, enabling us to assess effectiveness of phone call tutorials in this additional education emergency context. Results show that the typhoon was associated with reduced learning by approximately 0.12-0.20 standard deviations. Results also show that phone call tutorials continued to be effective, with 0.26 standard deviation gains relative to the control group, although SMS messages alone were not enough to stem learning losses. These results suggest that phone call tutorial effectiveness can persist across multiple types of education emergencies.

In addition to assessing impact on learning outcomes, we examine parent beliefs and demand for the intervention. We find parents and children both update their beliefs broadly in line with true learning progress. These results build on a literature exploring the ability to learn not only through being explicitly taught but also through noticing real-world progress (Hanna, Mullainathan, and Schwartzstein 2014) as well as a literature on the importance of parents knowing their child’s learning level, enabling them to better support their education (Bergman 2021). In addition, parent caregivers demand the program. At endline, 97 percent of parents state they would like to receive the phone call tutorials, which increases even more in the phone call tutorial treatment group, as does willingness to pay for the program.

We also examine impacts on non-cognitive skills, such as perseverance and ambition. We adapt a questionnaire used by Carlana and La Ferrara (2021). While we don’t find statistically significant effects for SMS messages on their own, we find sizable effects on both of these outcomes in the phone call tutorials, with up to a 29 percent increase in ambition. We also find positive effects on measures of well-being, such as enjoying school and worrying less. A growing literature highlights the importance of both cognitive as well as non-cognitive skills for future life outcomes (Jackson 2018). These results provide experimental evidence that some educational interventions can promote both types of skills (Alan, Boneva, Ertac, 2019).

A follow-on scale-up to the five-country randomized controlled trials provides an opportunity to evaluate effects at scale and in a post-pandemic setting. In 2024, the government of Karnataka, India announced that it had budgeted for a scale-up and planned to roll out the phone call targeted tutorial treatment with over 25,000 teachers, over an order of magnitude larger than the sample in our randomized trials. We analyze effectiveness using a difference-in-difference quasi-experimental strategy and find effects are substantial even at this much larger scale.

This study contributes to a few literatures. First, we contribute to a nascent experimental literature on education in emergencies. Substantial research has taken place on this topic, however much of it has been qualitative or with small samples. One exception is a randomized trial where

et al. 2019), versus cognitive skills. We find no statistically significant effects of the interventions on effort, revealing that learning gains are indeed a function of cognitive skills.

an NGO provided schooling in rural areas of conflict-affected regions in Afghanistan and found large effects on learning and closing of gender gaps (Burde and Linden 2013). We build on this literature by providing evidence from randomized controlled trials across multiple contexts and for government delivery models. We also expand the literature by evaluating alternative, scalable forms of education beyond traditional in-person schools, such as remote learning, which is the only option during many emergencies.⁵ A growing literature explores the effectiveness of remote tutoring, yet this literature remains mixed (Schueler and Rodriguez-Segura 2021; Crawford et al 2023; Gortazar et al. 2024; Carlana and La Ferrara 2024). Without a coordinated cross-country trial it is impossible to know if differences in effects are due to the intervention or to the setting, since both vary simultaneously. For example, Crawford et al. (2023) test a related but different intervention in Sierra Leone and find no effect of mobile phone tutoring. In their study, the tutoring call linked to a uniform national radio program rather than targeting instruction based on student learning week-on-week – a key difference. To assess whether the approach works across contexts we cannot simply aggregate prior studies where both intervention and context vary – we need to hold the intervention constant, while varying sites in a coordinated multi-site trial. We conduct such a coordinated five-country trial in this study.

We also contribute to a growing literature on external validity and scale. Recent examples show the extent of the scaling challenge, with many social programs which initially worked in proof-of-concepts no longer delivering impact when scaled or delivered by governments (Mobarak 2022; List 2024).⁶ In contrast with some existing studies, our results identify an education approach that appears to work well across contexts and when delivered by governments.^{7,8} We also advance the scale literature by introducing novel experimental evaluation both of government relative to NGO delivery and randomizing those teachers who implement to evaluate effects on ongoing teacher practices. This experimental variation enables us to assess questions of effectiveness within government systems. Moreover, we assess effects at scale with a quasi-experimental evaluation of a large-scale

⁵There is a robust literature now documenting the cost of school disruptions, due to teacher strikes (Jaume and Willen 2019), earthquakes (Andrabi et al. 2021), schools holidays (Cooper et al. 1996), and COVID-19 (Bacher-Hicks et al. 2021; Jack et al. 2021; Patrinos et al. 2022; Moscoviz and Evans 2022; Hevia et al. 2022; Carlana, La Ferrara, and Lopez 2023). However, less evidence exists on effective approaches to stem these learning losses. We contribute evidence on scalable solutions to stem learning losses across multiple contexts and through government delivery models.

⁶One example includes a contract teacher program in Kenya where effects dissipated when delivered by the government (Bold et al. 2018). Another example includes the diminishing effects of early childhood programs as they were scaled up from an efficacy trial (“proof of concept”) in Jamaica, to a pilot in Colombia, to an at-scale program in Peru (Araujo, Rubio-Codina, and Schady 2021).

⁷This is consistent with a set of studies which also focus on targeted educational instruction programs, and have also worked when delivered by governments (Duflo et al., 2020; Banerjee et al. 2017)

⁸One reason might be that we leverage a particularly scalable technology: mobile phones (Aker and Mbiti 2010; Aker et al. 2012). A second reason may be the generalizability of the underlying mechanisms tested in our study, which relate to other best practices in education. For example, tutoring has been shown to be one of the most effective, although expensive, approaches to improving learning in high-income settings (Nickow, Oreopoulos, and Quan, 2020; Robinson and Loeb 2021). The phone call tutorials provide a cheaper version of tutoring applicable in low- and middle-income contexts. In addition, since the phone calls are one-on-one and have frequent learning assessments, they enable targeted instruction to every child’s level, another approach shown to consistently improve learning (Banerjee et al. 2017; Muralidharan, Singh, and Ganimian 2019; Duflo, Kiessel, and Lucas 2020).

rollout by the government in India.

Our study also advances the external validity literature by conducting randomized trials across five countries in a literature where less than 1 percent of RCTs in a set of top economics journal articles are multi-country studies.⁹ Some argue that while RCTs address internal validity concerns, they might not address important external validity questions (Pritchett and Sandefur 2015). This study highlights the potential for randomized trials to be conducted and coordinated across contexts – addressing both internal and external validity challenges simultaneously. While rare, this approach is gaining traction, with another prominent example including a multi-country study on microcredit (Banerjee et al. 2015). This type of study, which tests effectiveness across settings, substantially enhances scientific reliability (Maniadis, Tufano, and List 2017). A prominent alternative approach to addressing external validity concerns includes adjusting for covariates and extrapolating treatment effects from a source setting to a target setting (Hotz, Imbens, and Mortimer 2005; Dehejia, Pop-Eleches, Samii, 2019). We conduct this exercise as a comparison, using results from a single setting and rich information on baseline covariates to extrapolate effects across sites. We then compare estimates with randomized trial estimates, a validation made possible by this multi-site study. We find extrapolation based on covariates, a leading approach to address external validity concerns, is not reliably able to predict causal estimates across settings.¹⁰ This result motivates more multi-site trials to reliably assess external validity.

Finally, we contribute to the global education literature, with a focus on improving learning outcomes. Over the past several decades, education enrollments have improved worldwide, yet learning has barely budged (Pritchett 2013; World Bank 2018; Angrist et al. 2021). Growing evidence reveals that popular input-only reforms, such as general teacher training, provision of computers, or school grants, are not enough to improve learning. In contrast, approaches that improve the quality of teaching, such as teaching at the right level and structured pedagogy, can generate large improvements in learning (Kremer, Brannen, and Glennerster 2013; Ganimian and Murnane 2016; Glewwe and Muralidharan 2016; Eble et al. 2021). The learning gains from the phone-based tutorials tested in this study can deliver up to 4 years of high-quality schooling per \$100 (Angrist et al. 2024). These types of highly cost-effective approaches can more easily scale and help address a persistent global learning crisis in low- and middle-income contexts.

⁹Out of a set of 400 papers in development from 2019 and 2021 in a set of top economics journals, 19 percent were RCTs; of those that were RCTs, about 1 percent were multi-country studies. The set of journals considered includes the Top 5 economic journals (American Economic Review, Quarterly Journal of Economics, Econometrica, Journal of Political Economy, and Review of Economic Studies) and other top-tier general interest journals (Review of Economics and Statistics, Economic Journal, Journal of the European Economic Association, and all four American Economic Journal journals), and a top field journal (the Journal of Development Economics). Other prominent multi-country RCT efforts include the evaluations of Graduation programs (Banerjee et al. 2015) and Teaching at the Right Level (Banerjee et al. 2017) and some early grade reading interventions (Lucas et al. 2014).

¹⁰In this setting baseline covariates are fairly similar across sites and we find limited treatment effect heterogeneity. Thus, when extrapolating effects from a given source setting, predicted effects only diverge somewhat from the source setting, whereas true treatment effects identified in our randomized trials show more variation.

II Conceptual Framework

In this section, we outline a framework with two key components underlying the effectiveness of phone-based tutoring: platform and pedagogy. In terms of platform, the program tested reaches students on an accessible, widely available platform: mobile phones. In terms of pedagogy, the program adapts a proven pedagogical approach: targeting teaching to a student’s learning level rather than their age or grade. We describe the framework further below, and prior evidence which informs each component. We come back to the conceptual framework later when discussing results in the mechanisms section to explore program effectiveness across contexts.

II.A Platform: reach at the right level

The phone-based tutoring program studied leverages the scalable and affordable technology of mobile phones. The program was designed to reach people ‘at the right level’ on a platform most households have easy access to.¹¹ In our five study countries, average mobile phone penetration was 108 mobile phone subscriptions per 100 people (ITU 2021). Coverage is also high, with at least 97% of the world’s population covered by at least a 2G mobile network, which is often sufficient to make and receive calls, and a minimum of 93% of households in our study countries.

Given high access to mobile phones, even in low resource settings, this platform presents a scalable way to deliver educational content to households (Aker and Mbiti 2010; Aker et al. 2012). While other technology platforms also present some opportunities for educational instruction, such as television, radio, and online media, access is often lower than through mobile phones. In low income countries, less than 30 percent of households report owning a television, 50 percent a radio, and 20 percent having internet access, suggesting content delivered over these platforms is less likely to have high uptake, particularly amongst the most disadvantaged households (CGD 2020). This is confirmed by households in our sample who report low usage of educational resources delivered via non-phone platforms (see Section III.D for details by country).

Phones also reduce the frictions that may be associated with some other platforms in emergency and low resource settings. For example, given phones are smaller than televisions, radios, and computers, they are often easily transportable in an emergency. Phones can also charge faster and last longer than computers and tablets. Finally, phone call interventions can also offer needed flexibility. While radio and TV programs often require caregivers to find dedicated time in their schedules, tuning into a pre-scheduled television or radio program, phone call programs can provide more flexible scheduling options, increasing take-up. Households receive calls from a tutor who solicits their interest in participating, then initiates the tutoring calls and schedules according to a given household’s availability, imposing minimal coordination costs on participating households. Moreover, since the tutor calls the household, there is minimal to no cost to the household to participate.

¹¹We thank Rukmini Banerji, CEO of Pratham, who inspired the use of the term ‘reach at the right level.’

In summary, a key mechanism through which phone call programs may improve learning outcomes is by reaching households ‘at the right level’ on a platform most households can easily access.

II.B Pedagogy: teach at the right level

There is a growing literature on the effectiveness of targeting teaching instruction to a student’s level rather than their age or grade (Banerjee et al., 2017). The principle of targeted instruction focuses on identifying each student’s individual learning level through a light-touch learning assessment. Instructors then use this information to target their instruction and ‘teach at the right level.’ Targeting instruction stands in stark contrast with business-as-usual teaching in most education systems. In most systems, teachers typically teach to a one-size-fits-all grade-level curriculum. Yet most children are not at grade level. For example, while grade-level curricula often expect students in grades 3 to 5 to be able to do two-digit division, in our study sample, only 7 percent of students can do so. This phenomenon is pervasive. A survey in Kenya, Tanzania, and Uganda showed that three-quarters of students in grade 3 could not even read a simple sentence, falling well below grade-level expectations (World Bank 2018). In this context, students being taught grade-level material will be left behind and stay behind. In contrast, targeting instruction to students learning levels, focusing on foundational skills, via rapid assessment and teaching activities tailored to their level, enables students to learn far more effectively.

Targeted instruction approaches have received growing attention as highly cost-effective relative to status quo teaching and have been tested using various in-person and in-school models (Banerjee et al. 2017; Muralidharan, Singh, and Ganimian 2019; Dufflo, Kiessel, and Lucas 2020; Angrist and Meager 2023). The phone call tutoring program tested in this study adapts the targeted instruction pedagogy for a remote setting, with frequent assessments conducted weekly to enable ongoing, flexible targeted instruction even in disrupted settings. In addition, phone call tutorials mimic some of the targeting benefits of tutoring programs, which are conducted in smaller groups, thus also enabling more targeted instruction to each child’s level. Tutoring is an approach most often tested to date in high-income settings (Nickow, Oreopoulos, and Quan, 2020; Robinson and Loeb 2021). We adapt tutoring approaches to low- and middle-income settings, testing a low-cost phone call tutorial model targeted to student levels and made affordable via use of mobile phones.

The two components outlined in this framework – platform and pedagogy – come together to both reach at the right level as well as teach at the right level. This framework builds on prior evidence, and provides conceptual clarity on the mechanisms underpinning the potential effectiveness of phone call tutorials across diverse and disrupted contexts.

III Context

III.A A global learning crisis, exacerbated by COVID-19

While schooling rates have increased worldwide, students face a global learning crisis, with many students in school but learning very little (World Bank 2018). In our baseline survey, for example, 36 percent of students in grades 3 to 5 could not do any basic operations, falling well behind curriculum expectations. COVID-19 school closures exacerbated this global learning crisis, with school closures forcing over 1.6 billion learners out of classrooms. These types of school closures result in large learning losses which have been documented in North America, Western Europe, Asia, and Sub-Saharan Africa. The lasting economic toll of learning loss is enormous, estimated at over 20 trillion dollars (Azevedo et al. 2020). This pre-existing and exacerbated learning crisis necessitates effective educational approaches that can rapidly improve learning outcomes during and beyond emergencies.

III.B Education in Emergencies

COVID-19 school closures are a recent large-scale example of disruptions to schooling. However, emergencies that affect education are all too common, and affect many people every year. We present novel data documenting the extent of school closures over the past two decades for a subset of countries. We codify information from various sources, including reports from international aid organizations such as Save the Children and UNICEF, the United Nations Office for the Coordination of Humanitarian Affairs (OCHA), as well as national and international news reports.

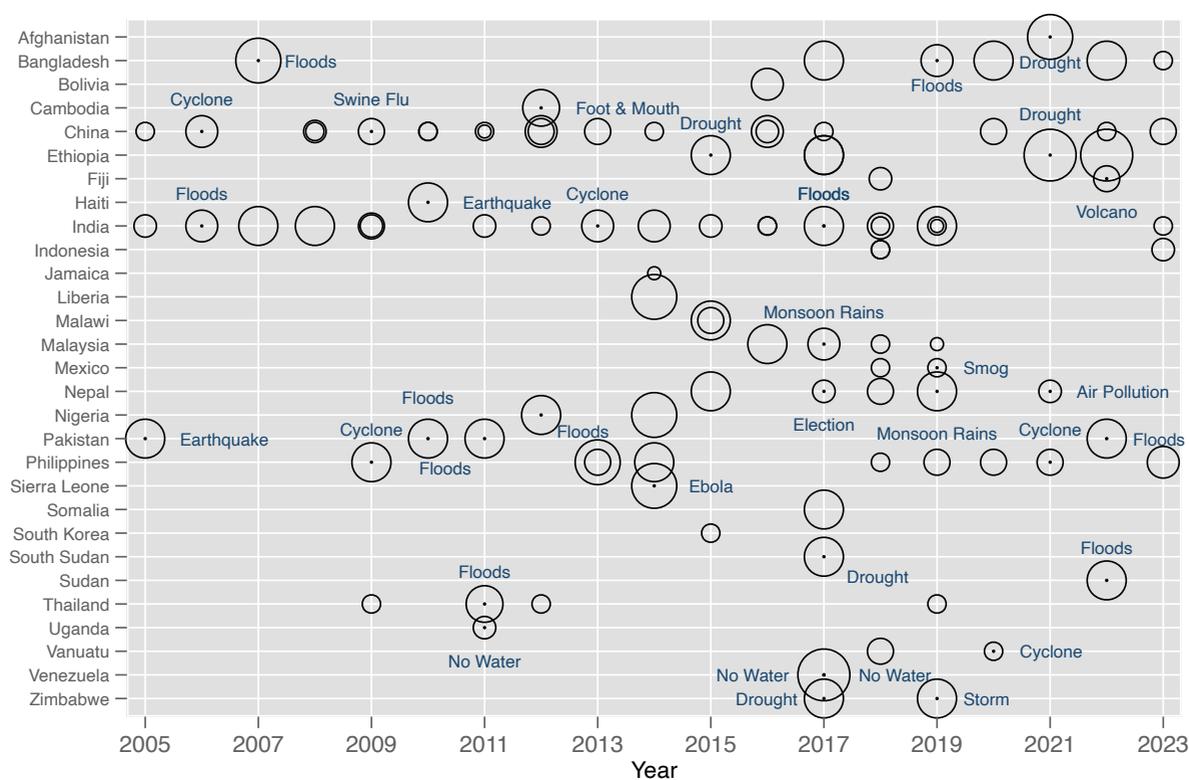
Figure 1 plots an index of the length of school closures and number of people affected by shocks that have disrupted schooling by country and year. We observe that large-scale disasters affect many people, occur in many countries, and happen at high frequency. The spread of multiple diseases has shuttered schools, including Swine Flu in China in 2009, Foot and Mouth in Cambodia in 2012, and Ebola in Sierra Leone in 2014. In 2019, smog and air pollution from wildfires closed schools around Mexico City. In 2020, a cyclone in Vanuatu destroyed and damaged 885 schools. Extreme climate events often disrupt schooling, such as water shortages in Venezuela in 2017. Floods and rainy seasons perhaps most routinely cause widespread school disruption. For example, in 2017, monsoons and flooding closed and damaged over 10,000 schools across Malaysia, India, Bangladesh and Nepal. In Pakistan, nearly 8,000 schools were damaged in 2010. A more recent crisis in Pakistan in 2023, which will be included in an extension of this dataset, disrupted 34,000 schools. As climate change-related shocks increase, school disruptions are likely to occur with ever great frequency and consequence, stunting students' education.

In recognition of the importance of addressing shocks to education, the United Nations established a billion-dollar Global Fund called "Education Cannot Wait" in 2016 at the World Humanitarian Summit to support learning in emergencies and protracted crises. Multiple additional international aid structures also exist to support education in emergencies. For example, the Inter-

agency Network for Education in Emergencies (INEE) was established to connect and share best practices amongst more than 18,000 members from 4,000 institutions globally working in education in emergencies, including the United Nations, governments, NGOs, teachers, and students.

A prominent review by Burde et al. (2015) on education in emergencies notes that most evaluations in the sector have been qualitative. The review highlights a major gap in experimental studies and the need for interventions which focus on improving education quality, in addition to education access, in emergency settings. They identify the use of mobile phone technology to deliver educational instruction as a particularly promising yet underutilized approach, and recommend future experimental research in this area. This study aims to address these critical research gaps, contributing experimental evidence on education in emergency programs which can promote learning cost-effectively, across settings, and with governments.

Figure 1: Documenting a Set of Education in Emergencies (2005-2023)



Notes: This figure plots an index of the length of school closures and number of people affected by shocks that have disrupted schooling by country and year. The larger the bubble the larger either the length of school closure or the number of people affected, or both. More information on the database compilation is available in the Appendix.

Source: Compiled school closure information based on press releases of the United Nation's Office for the Coordination of Humanitarian Affairs (OCHA) ReliefWeb, World Vision, UNICEF, the BBC, and other local outlets. Major disruptions were also identified through the international disaster database (EM-DAT).

III.C An Overview of the Study: Randomized Trials Across Five Countries

Our evaluation of education in emergencies interventions took place in the context of COVID-19 school closures across five countries: India, Kenya, Nepal, the Philippines, and Uganda. In total, over 16,000 households were enrolled in this multi-country evaluation. In each household, a primary school student and a caregiver were identified to receive the intervention. We asked households to nominate the main caregiver who provides educational support to their child: 53.3 percent of nominated caregivers were mothers, 17.4 percent were fathers, 17.3 percent were siblings, and 3.2 percent were grandparents. 24 percent of nominated caregivers had completed primary education or less. Students were enrolled in grades 3 to 5, with one exception in Kenya where students were in grades 1 and 2. The trials took place between December 2020 and July 2022. Appendix Table A8 includes a summary of key facts per country trial, and Figure A3 in the Appendix shows a detailed timeline. Appendix Figure A6 maps the location of the children’s schools within several of the study countries; it shows study participants are distributed across wide areas in most countries, and in some cases, are relatively nationally representative. Students provided phone numbers where they could be contacted, which was most often their parent’s phone number, and in some rare cases, the phone of a neighbor or other community member.

Our evaluation includes multiple education in emergencies programs and delivery models. We leverage mobile phones – a high-access, low cost and scalable way to reach students and their caregivers when school is out of session – to provide various educational interventions over the course of eight weeks.¹² One treatment includes a set of SMS messages, such as numeracy content provided weekly, as well as nudges to engage in educational activity. A second treatment added weekly one-on-one 20-minute phone call tutorials. Phone calls covered foundational numeracy and aimed to target instruction to student learning levels via light-touch, high-frequency assessments conducted on a weekly basis. Using the assessment, instructors led students through foundational numeracy practice problems to help them master addition, subtraction, multiplication, and division. For example, students who did not know addition would be taught addition, while students who knew addition but not subtraction would be taught subtraction. Calls were made to caregivers who invited the child in their household to join the program, and would then hand them the phone or use speakerphone for the tutorial session. The implementation ratio of students to teachers was approximately 20:1 for teachers implementing the program full-time, and 5:1 for teachers implementing the program part-time. The phone call program is called “ConnectEd”, highlighting phone calls’ ability to connect students to quality education even during school disruption.

In addition, we tested several scalable delivery models, including NGO and government teacher delivery, which were randomized within a country. The government delivery arms were conducted with government teachers. They included headquarter ministry sign off, as well as regional government engagement, including joining training, engagement with routine monitoring data, and

¹²These types of mobile phone-based interventions have shown to be able to close parent-child information frictions and improve learning in high-income settings (Bergman 2021) as well as low and middle-income settings (Angrist, Bergman, and Matsheng 2022).

periodic mid and high-level steering committee meetings. Some delivery models hired teacher aides to deliver instruction, while others used teachers.¹³

We include a few descriptive statistics to describe our samples. In our five-country sample, the median student is in grade 3 and 51 percent of students are female. On baseline student learning proficiency, 36 percent of students did not know any basic operations, 27 percent were at addition level (meaning they had ‘mastered’ addition and were being taught the next level), 12 percent were at subtraction level, 17 percent at multiplication level, and 7 percent at division.¹⁴ These low learning levels are well below grade-level curriculum expectations. Moreover, detailed monitoring data shows high week-on-week engagement across sites ranging from 70 to 80 percent, revealing the scalability and robustness of the approach even in disrupted and low-resource environments, as shown in Figure A1. We also find sustained rates of engagement across all eight weeks of phone call tutorials staying within a narrow range of high engagement.

This set of five trials sought to scale an approach first tested in Botswana in 2020 (Angrist, Bergman, and Matsheng 2022). The original Botswana trial tested the impact of a phone based tutoring program that had been developed during the first few months of COVID-19. The NGO Youth Impact, one of the largest in the country, developed a phone call that targeted tutoring for remote settings when schools were disrupted. The Botswana study – a tightly controlled initial proof-of-concept – found the program to be cost-effective, resulting in a 0.12 standard deviation improvement in learning outcomes. The present study sought to test the scalability of the approach across five country contexts and a variety of implementer types, including scalability through government delivery models. A rich array of data was collected with response rates above 80 percent, enabling both high-quality evaluation and detailed exploration of outcomes and mechanisms.

In all contexts, the NGO that first developed the program, Youth Impact, provided support to each of the implementing partners in new countries to help train their staff in the program, as well as provide monitoring tools, training, and technical assistance. Support typically involved providing the curriculum and advice if any adaptations were needed for the local context (for example, these ranged from careful language translation to adapting material if place value was taught differently in new contexts); a several-hour training-of-trainers session; monitoring surveys and advice on how to use the monitoring data tools; and occasional meetings during implementation to give advice and troubleshoot. All interactions between the research team and implementing partners took place remotely over video or voice calls. Training of trainers was led by a Youth Impact master trainer for each local implementing partner, who in turn directly trained teachers and tutors in their context and language. Training was practical and interactive, with breakout groups to practice tutoring delivery, followed by live phone calls to practice with real households.

Once implementation began, each week, program managers in each implementing organisation

¹³Note: we use the terms ‘teacher’ and ‘teacher aide’ to reflect several types of trained teaching instructors. Those referred to as ‘teachers’ are currently employed primary school teachers. ‘Teacher aides’ include community education volunteers as well as instructors trained to be teachers but currently employed by NGOs. Details of implementer types are discussed in each country section below.

¹⁴This excludes Kenya, which is the only country where we do not have baseline student level data identical to the other countries broken down by specific proficiency, though we have baseline test score data on a 100-point scale.

collected and analysed weekly monitoring data submitted by all tutors. This allowed the program coordinators in each implementing organisation to monitor and follow up on the weekly phone call success rate and program delivery. Implementing organisations used this data to provide support to help teachers and tutors troubleshoot, reach more students, improve targeting accuracy, and share lessons and tips across tutors.¹⁵ Youth Impact supported partners across countries, providing a coordinating mechanism to share lessons across trials. However, Youth Impact had no prior presence in these countries, showcasing a replicable model for scale-up in new settings.

A typical phone call involved the following format. The tutor would call the household after arranging a mutually agreeable time. Phone calls were directed toward the student, although caregivers were encouraged to place the phone on speaker and be available to support. Working at the student’s identified learning level from their previous week’s performance, tutors guided students through a simple set of 3-4 steps on how they should solve the operation being taught that week (addition, subtraction, multiplication, or division). Once the student had walked through a series of problems with the tutor and they had solved a few problems together, then the call concluded with a ‘checkpoint question’. This consisted of one math question at the level taught that week. This question enabled tutors to evaluate and update the child’s learning level to ensure they could target instruction to the child’s correct level the following week. For example, a student who could not do the week’s addition checkpoint question would receive instruction in addition in the following week’s call; if they got addition correct, the following week they would be taught subtraction.

III.D Specific Country Context for Each Trial

For each country context, we describe the status quo learning levels and COVID-19 education context, the implementing partner and sample description, and the instructor type. Figure A8 characterizes school disruptions in all five countries. In each setting, learning gaps are large both before and during COVID-19, showing a need for more high-quality and scalable educational instruction. In addition, the typical government responses to school disruption, such as radio and TV, have engagement rates below 30 percent.¹⁶ This highlights the need for remote approaches which can yield higher engagement, such as phone call tutorials. Moreover, this puts in context the significance of the high engagement we find of the phone call tutorials in our study, consistently reaching over 70 to 80 percent. Finally, the diverse settings and implementation models in the study, spanning five countries as well as NGO and government delivery, enables us to assess the scalability of targeted phone tutoring across a wide array of contexts and delivery models.

¹⁵See details in Appendix D for additional information on training and implementation.

¹⁶In all contexts, support by governments and school districts during disruption was mixed, with provision of some public education support such as television and radio programming, and encouragement of teachers to support their students remotely, though evidence suggested limited uptake of these services. In this section we outline what services were available and taken up in the countries where we have data.

III.D.1 India

Prior to COVID-19, India placed near the bottom of all countries taking the Program for International Student Assessment (PISA), ranking 72nd out of 73 countries. In our baseline survey, only 24 percent of the Grade 3-5 students in the sample could do division, showing a substantial need for additional foundational numeracy instruction.

During COVID-19, schooling was significantly disrupted in India. In our sample in Telangana state, primary schools were partially closed for the duration of the program (UNESCO 2023). In response to the crisis, the government encouraged households to access educational materials shared on TV, online, and radio.

The implementing NGO in India was Alokita, a non-governmental organisation based in Hyderabad, which is part of an international network of organizations called Global School Leaders. Alokita provides education and school leadership-related programming in schools. The program supported government efforts to ensure all primary school students achieve foundational numeracy skills.

Alokita had pre-existing relationships with government officials and study schools, nearly all of which were government residential schools. The majority of the students were from rural areas and belong to India's most marginalised caste communities. Between two to four teachers were selected from each school to tutor the students. The phone calls were delivered by school teachers based in the same schools as the students in the study. Calls were conducted in either of two languages, English or Telugu, the most common language in the state.

III.D.2 Kenya

Although Kenyan students have higher foundational learning levels than most other countries in Sub-Saharan Africa, learning outcomes still lag far behind those in higher-income countries (World Bank 2020). In third grade, only 47 percent of Kenyan students could solve a second grade mathematics problem (Uwezo, 2016), and 36 percent of grade 2 and 3 students achieved a minimum proficiency level in mathematics (UNESCO, 2016).

By December 2020, when the study launched, schools had been fully closed for 9 months. Schools re-opened from early 2021 onwards, partway through the intervention. The main distance learning programs available to students spanned TV, radio, and online and phone-based educational offerings.

The study took place in 30 of Kenya's 47 counties, in the universe of 112 schools the partner operated. NewGlobe, the partner NGO of the Kenya study, operates one of the largest, low-cost private school networks across the country. In Kenya 33 percent of pre-primary students are enrolled in private schools, and 16 percent of primary students (UNESCO 2022). Kenya was the only context where grades 1 and 2 were included in the program, rather than grades 3 through 5. This was done because the baseline level of foundational numeracy skills was higher than in other contexts, and thus earlier grades were most comparable in terms of baseline learning levels.

The phone call program was implemented by students’ normal class teacher, who undertook the phone calls during school closures, as well as after school as part of their teaching duties once schools re-opened part-way through the study. Randomization was at the teacher level, equivalent to the school-grade level. Teachers were advised to deliver tutoring sessions in English, and also conducted some parent interactions in Kiswahili.

III.D.3 Nepal

In 2018, only 28 percent of Grade 5 students in Nepal demonstrated grade-level proficiency in mathematics (NASA 2018; Radhakrishnan et al., 2021). In our baseline survey, only 5 percent of the Grade 3 to 5 students in the sample could do division.

Nepal’s schools were closed when the trial started in January 2021. While partial re-openings occurred mid-way through implementation, school disruptions (e.g. frequent closures and re-openings in response to COVID-19 infection waves) were common throughout the study. During this time, the government rolled out learning programs using radio, television, and online sources, and disseminated printed learning materials to students. However, few students were able to access existing remote educational support materials during school closures. Baseline data showed that in the status quo, only 31 percent of students had teacher interactions during school closure and less than five percent accessed radio or online education (Radhakrishnan et al., 2021). These gaps in remote learning access highlight the need to provide additional support, such as ongoing teacher phone calls to tutor their students.

The study sample included 10 local governments selected by the Ministry of Education, Science and Technology (MoEST) and World Bank, and covered all 7 provinces, with broad geographic spread in the country. The sample included public school students in grades 3 through 5.

To assess scalability, the phone call tutorials were implemented through both a government delivery treatment arm and an NGO delivery arm. Students were randomly assigned to either arm or a control group. The program was implemented by a coalition of partners including MoEST, local governments, the World Bank, Teach for Nepal, and Street Child. In the government arm, public school teachers implemented the program, with teachers teaching students from outside of their own region.¹⁷ In the NGO arm, facilitators trained as teachers delivered the phone call tutoring program. Calls were conducted in over 12 languages, with students matched to teachers who spoke their language.

III.D.4 Philippines

The Philippines scored second-lowest of 79 countries in the 2018 PISA mathematics assessment. During COVID-19, Philippines students also faced one of the world’s longest school closures, with schools closed for over two years (UNESCO 2023). Only 2 percent of students in our baseline

¹⁷In an additional randomised trial discussed below, we used the over-subscribed list of eligible public school teachers in Nepal to randomly assign teachers to an implementation ‘treatment’ arm, or control arm. We then also measured causal effects of being an instructor on teacher beliefs and practices.

sample could do division, placing them far behind grade-level expectations; these baseline learning levels are some of the lowest in our study.

At the start of our Philippines study in August 2021, schools had been closed for almost 1.5 years, and they remained closed throughout the trial. During school closures, the government encouraged households to access educational content provided on radio, television, online, and printed materials. However our data reveals limited use of distance educational materials, with fewer than 5 percent accessing radio and TV resources. These gaps, like in other countries, reveal the need for additional high-quality educational support.

The program was implemented in 3 of the country’s 17 regions. These regions were selected by the government and represent some of the most marginalized communities in the country. Students from grades 3 and 4 were enrolled in the study.

In the Philippines, like Nepal, the phone call tutoring program was implemented through a government delivery treatment arm and an NGO delivery arm. In the government arm, public school teachers employed by the Philippines Department of Education implemented the program. Working with the government, Innovations for Poverty Action, a research NGO, trained government teachers from participating schools on the intervention. The teachers were assigned to call students at their school in the grade they taught. In the NGO-led teacher-aide arm, Innovations for Poverty Action in coordination with the central and regional government offices, hired and trained tutors from the pool of government teacher applicants to implement the program. Calls were conducted in the 5 most commonly spoken languages in the study regions.

III.D.5 Uganda

Ugandan students are well below grade-level expectations in foundational skills. Recent data shows that just 6 percent of grade 4 students are able to read a paragraph and only 2 percent can solve a simple math problem (World Bank 2019). On our baseline survey, only 14 percent of the Grade 3-5 students could do division.

The learning crisis was compounded when, like the Philippines, Uganda entered one of the world’s longest COVID-induced school closures, with schools closed for nearly two years from March 2020 to January 2022 (UNESCO 2023). While the government provided distance learning support to households, engagement with these materials was limited (CESS 2021). Only 29 percent of households engaged in radio lessons, 22 percent in printed self-study materials, and under 12 percent in TV and online (Uwezo Uganda, 2021). Like in other contexts, this highlights the need for higher-engagement approaches, such as live phone calls.

The study took place in 9 out of Uganda’s 135 Districts, which covered some of the most rural part of the country. The sample included students in grades 3 to 5.

In this study, the implementing organisation was Building Tomorrow, an education-focused NGO in Uganda. Phone calls were made by Building Tomorrow’s Community Education Volunteers, resident members of communities who are recruited and trained to serve as grassroots education ‘extension agents’. These volunteers encourage out-of-school children to enroll in school,

and lead literacy and numeracy lessons in community settings. In Uganda, calls were conducted in one of three languages: English, Luganda, and Runyankore.

IV Data

In each country, we have two waves of data: baseline and endline.¹⁸ The endline surveys took approximately 30 minutes to administer and included approximately 20 questions. These questions included a learning assessment, child wellbeing, parental engagement in educational activities, and parental perceptions of their child’s learning. A portion of the survey was conducted with the parent, and learning outcomes were collected by directly assessing the child over the phone. Endline surveys were conducted a few months after the program ended. A set of common core questions in the baseline and endline surveys are included in Appendix C.

The learning assessment was adapted from the ASER test, which has been used frequently in the literature to measure learning outcomes (Banerjee et al. 2007; Banerjee et al. 2017) and is used routinely across 14 countries. The test consists of multiple numeracy items, including two-digit addition, subtraction, multiplication, and division problems. In addition to the ASER test, we asked students to solve a place value word problem, a fraction problem, and an addition word problem to capture learning outcomes beyond a core set of mathematical operations.

To maximize the reliability of the phone-based assessment, we introduced a series of quality-assurance measures. To minimize the likelihood of family members in the household assisting the child, students had a time cap of two minutes per question and we asked each child to explain their work. We only marked a problem correct if the child correctly explained how they solved the problem and enumerators were confident parents were not assisting their child. We also conducted a battery of validity checks to ensure the reliability of the learning outcomes. A first robustness check compares in-person to phone-based assessment for the exact same set of students.¹⁹ An additional test included back-checks, with a random subset of students tested twice on the same competencies. We also randomized various problems of the same proficiency (e.g. four different questions to measure 2-digit addition with carryover). Finally, we include a real-effort question to disentangle effects of the intervention on effort on the test versus cognitive skills. Prior research suggests effort on the test can affect test scores during in-person exams (Gneezy et al. 2019). Thus, by providing a measure of effort, we can disentangle effort effects versus cognitive skills gains.

The survey also included questions on caregiver engagement in their child’s education and beliefs. We measured engagement by asking caregivers how often they spent helping their child with their schoolwork over the previous weeks. We also included a measure of a caregiver’s confidence in their child having made progress in learning over the previous months, and their perception of

¹⁸In Nepal, we conducted a baseline survey with a random 50 percent of the sample, and in Kenya, we relied on administrative data instead of baseline survey data.

¹⁹This quality assurance test was conducted in Kenya. All study students first completed the phone-based endline assessment, followed by the in-person endline assessment. Tests asked identical questions, assessing performance on the same concepts, question types, and difficulty level. This enables us to assess how test administration method affects scores and to include standard in-person assessments in the analysis.

their child’s numeracy level. Additional questions included information on whether the caregiver has returned to work. We also asked about parents’ demand for remote learning services in the future, and whether they would be willing to pay for such a program. For students, we asked about child’s mental wellbeing, how much they enjoy school, and the child’s own belief about what math problems they will be able to answer. We also measured non-cognitive skills, such as perseverance and ambition. Finally, we included demographic questions, recording the child’s age, grade, and gender.

The overall sample size, pooling all sites, is 16,936 households. For endline surveys, we randomly sampled households from the full sample to interview. This was due to time and cost constraints. In Appendix Table A1, we show that those randomly selected for endline interview are statistically equivalent to the full sample at baseline along a series of indicators such as gender, student grade, and baseline learning level. The samples randomly selected to be interviewed for endline yield a total endline sub-sample of 12,707.

Appendix Table A2 presents the response rate to the endline surveys and an analysis of survey attrition for those randomly selected to be part of the endline sample. The follow-up rate was very high at around 80 percent of respondents at endline. Table A2 also presents a test of whether response rates differed by treatment assignment. We find no evidence of differential response rates between treatment and control groups. We also find no evidence of differential attrition in any individual country. This provides evidence that our sample has a high and unbiased response rate.

Finally, we also include a survey for teachers to assess their beliefs and instructional practices. These questions include their desire to be a teacher, and teachers’ view that the phone call tutorial program was helpful for student learning. Questions also include instructional practices, such as involving parents in education further and better targeting feedback to students’ actual learning level. These questions can potentially capture persistent effects on educational systems through teachers changing their beliefs and behaviors beyond the lifecycle of the program.

V Experimental Design and Empirical Strategy

We ran randomized controlled trials (RCTs) across five countries. Every trial had a control group and a combined phone call and SMS treatment arm. All studies except the trial in India also included an SMS-only treatment arm. In two countries, Nepal and the Philippines, we further randomized delivery of the phone call by NGO instructors or government teachers to assess scalability within government systems. We exploit random assignment to identify causal effects and estimate the impact of these education in emergencies interventions and various scalable delivery models. We first estimate intent-to-treat effects as follows:

$$Y_{ij} = \alpha + \beta_1 \text{PhoneCalls}_j + \beta_2 \text{SMS}_j + \gamma X_j + \delta_s + \eta_c + \epsilon_{ij} \quad (1)$$

where Y_{ij} is a learning outcome for individual i in household j . PhoneCalls_j and SMS_j take on the value 1 in their respective treatment arms and 0 otherwise. X_j denotes a vector of baseline

control variables to enhance statistical power and precision.²⁰ δ_s refers to relevant strata in each study, which includes baseline learning, gender, school, and region fixed effects.²¹ We both assess impact on specific learning proficiencies as well as on standardized learning outcomes (standardized relative to control group standard deviations at endline and centered at mean zero). Distributions of raw baseline and endline learning levels are shown in Appendix Table A5. Our main specifications pool across studies and we include country fixed effects η_c . We estimate the regression using ordinary least squares (OLS). We run secondary regressions using alternative weights by country as well as country-by-country estimations.

We also ran a subset of trials where phone calls were delivered by NGOs or government teachers. We estimate government versus NGO delivery effects using the following specification:

$$Y_{ij} = \alpha + \beta_1 \text{PhoneCalls}_j \text{NGO} + \beta_1 \text{PhoneCalls}_j \text{Gov} + \gamma X_j + \delta_s + \eta_c + \epsilon_{ij} \quad (2)$$

In the main analysis of the pooled sample, no adjustments are made to reflect the differences in sample sizes between countries, so every observation is weighted equally. This follows standard practice in the analysis of multi-site RCTs. We test robustness using regressions that instead weight each country equally, and find similar results.

All standard errors are clustered at the level of the unit of randomization. In most countries this was the household level, which was de facto the same as the student level since one student participated per household; in a subset of treatment comparisons it was the school-grade level.²² The above specifications are the core estimation strategies. Additional specifications include other outcomes of interest beyond learning, such as engagement, as well as regressions run country by country, heterogeneity analysis, and various types of pooling across countries.

Due to randomization, we expect treatment and control groups to be balanced at baseline in expectation. Appendix Table A3 presents balance tests using baseline data for key demographics and the same learning level variable. We find no statistically significant differences across multiple dimensions, including gender, grade, caregiver type, caregiver education levels, and baseline learning. This is robust to multiple specifications, including with and without country fixed effects. Similar results are found for each country, with balanced samples in each individual country. This reveals that each randomization yielded balanced arms in line with expectations.

In all sites, the experimental design had high compliance rates; nearly all randomized treatment

²⁰These variables include: student grade and baseline learning levels

²¹The strata used in each trial were as follows. India: baseline level, gender, school; Kenya: school; Nepal: region, baseline level; Philippines: baseline level; Uganda: baseline level, school type; previous education program participation.

²²In India, Nepal, and Uganda all arms were randomized at the household level. In Kenya and the Philippines some arms were randomized at the household level (e.g. the NGO arms of the Philippines) and some at the school-grade level (e.g., the Government arm of the Philippines) with corresponding control groups randomized at the same level for each arm. In Kenya, the phone call arm was randomized at the school-grade level while the SMS arms were cross randomized at the household level. Of note, robustness checks find little to no difference if standard errors are clustered or not across all specifications. This is largely due to the fact that there is substantial variation in learning outcomes within a cluster, such that the intracluster correlation is very low at about .03 in the Philippines and Kenya.

households participated in the expected treatment group. Detailed weekly monitoring data shows high take-up and fidelity, as shown in Figure A1.

In a secondary analysis, we embed another randomized trial which randomly allocated a subset of government teachers to deliver the phone call tutorials out of a pool of hundreds of eligible teachers in Nepal. Appendix Table A4 shows balance tests with no statistically significant differences among treatment groups. We assess whether participating in phone call tutorials changed teachers’ subsequent beliefs and behaviors, which we estimate for teachers i at the school level j as follows:

$$S_{ij} = \alpha + \beta_1 TeacherCalls_i + \gamma X_i + \delta_s + \epsilon_{ij} \quad (3)$$

where S_{ij} captures teacher beliefs or instructional practices. These outcomes measure potential spillovers into the education system through teacher beliefs and behaviors which could persist beyond the program.

VI Results

VI.A Main Results: Learning Across Contexts and Scale Delivery Models

Our results show consistently large and robust effect sizes of phone call tutorials on learning across contexts, with average effects across all five countries between 0.30-0.35 standard deviations. Column 1 in Table 1 shows that for our main learning outcome of foundational numeracy skills—measured using average student level — we find large, statistically significant learning differences between treatment and control groups. For the combined phone and SMS group, there was a 0.327 standard deviation ($P < 0.001$) increase in the average numerical operation across all 5 countries. For the SMS messages group, we find smaller but still statistically significant improvements in learning, of 0.083 standard deviations ($P=0.003$). For both the SMS group and the combined Phone and SMS group, the magnitude of the impact and statistical significance are robust to different estimation approaches, for example including baseline controls and country and grade fixed effects (Column 2), and weighting results by country-arm (Columns 3 and 4).²³

These results are large relative to the typical education intervention’s effectiveness. A recent review by Evans and Yuan (2022) found that the median effective intervention yielded 0.10 standard deviation gains in learning. Moreover, a review of 150 interventions by Angrist et al. (2020) found that over half of education interventions do not work at all. These reviews put our results in context, with phone call tutorials three times as effective as the median education intervention. Moreover, effects are even larger in these multi-country studies than in the Botswana proof-of-concept study, where learning improved by 0.12 standard deviations (Angrist, Bergman, and Matsheng 2022). This result contrasts with prior literature showing that proof-of-concept studies rarely scale successfully to new contexts (List 2022) or experience diminishing returns (Caridad, Rubio-Codina, and Schady

²³We also adjust for multiple hypothesis testing (List et al. 2019). Table A10 presents the main learning outcomes and includes sharpened q-values to adjust for multiple hypothesis testing (Anderson 2008). Estimates remain highly statistically significant.

2021). Rather than finding diminishing returns, we find results improve as the approach is adapted, scaled, and tested across contexts. We explore potential explanations later in the paper, including learning from experience as well as higher need for the intervention, with longer and more disruptive school closures in the multi-country trials with up to two years of school closure.

In Table 2, we show results for the subset of countries (Nepal and the Philippines) where we randomized delivery models to test scalability within government systems. We compare government teachers with NGO delivery. Column 1 shows the pooled results: both government teachers and NGO instructors are effective at improving student learning, with government teachers improving learning by 0.314 standard deviations across both contexts, and NGO teacher aides improving learning by 0.263 standard deviations ($P < 0.001$ for both). The similarity in effect, with no statistically significant difference between models, demonstrates that these education in emergencies programs can be effectively implemented by government teachers.

These results show striking effectiveness when delivered by governments, on par with NGOs. Prior literature has found that programs often work when delivered by NGOs but fail to replicate when scaled by governments. For example, in Kenya, contract teachers improved learning, but when delivered by the government, program effectiveness waned (Bold et al. 2018). A large-scale government teacher training program in Nepal similarly found no effect, largely due to poor implementation (Schaffner, Glewwe, and Sharma 2021). Our results build on this literature, providing an alternative view: government delivery can achieve large learning gains. This is consistent with other studies which have refined interventions over time to facilitate government adoption (Banerjee et al. 2017; Duflo et al. 2020). Figure 2 visualizes results from Tables 1 and 2, with average effects across studies and split by government versus NGO delivery models.

We embedded an additional randomized study allocating a random subset of teachers to deliver the phone call tutorials in Nepal out of a list of hundreds of eligible government teachers.²⁴ This experimental variation enables detection of the impact of delivering the program on teachers' beliefs and practices – a measure of potential spillover effects into the education system that can persist through teachers beyond the intervention. In addition, teacher randomization enables generalization to the broader set of eligible government teachers. Thus, results capture government delivery impacts which do not hinge on implementers being highly selected, and effects are likely to translate to typical government teachers in the education system.

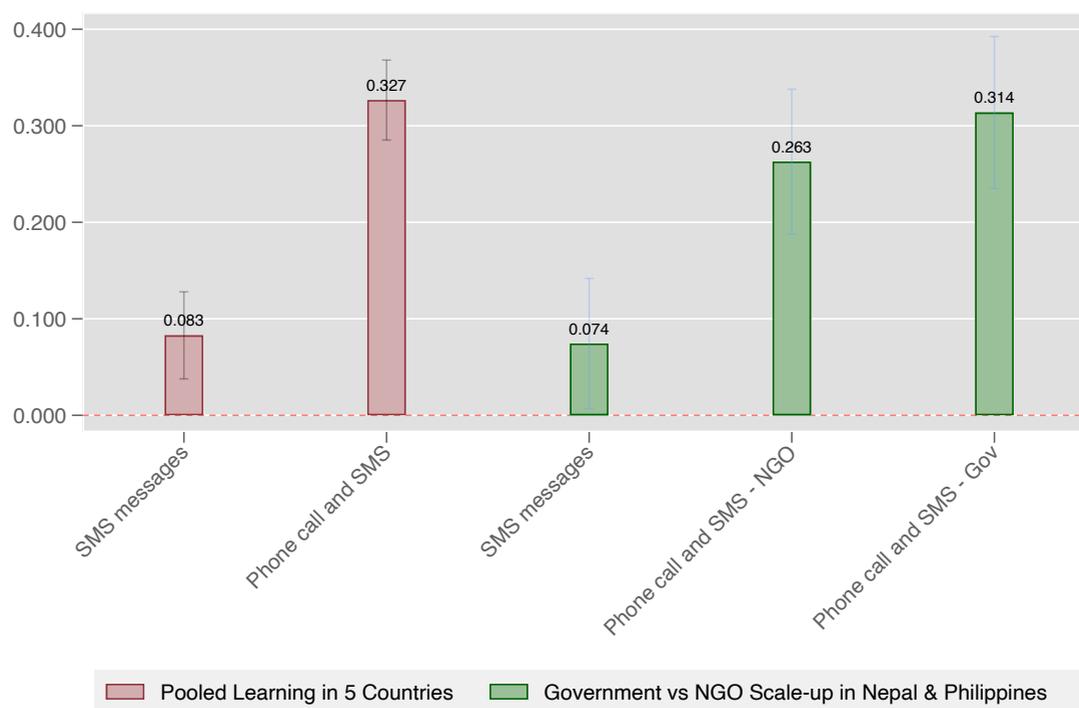
Table 3 shows teacher practices shift substantially; teachers are 9.3 percentage points more likely to target their feedback to students learning level ($P < 0.05$). Teachers are also more likely to get parents involved in education ($P < 0.1$). We further find large effects on teacher perceptions that they were able to help students learn, as well as their desire to teach, with a 15.8 percentage-point gain in wanting to be a teacher if they could make the choice again ($P < 0.1$ and $P < 0.01$ respectively). These results suggest that delivering effective programs could unlock a virtuous cycle within government education systems, in turn motivating teachers to want to teach and to improve

²⁴Over 80 percent of eligible teachers wanted to engage in the phone call tutorials. Of those enrolled in the study, 97 percent of both control and treatment teachers stated that they wanted to deliver the program in the future.

their teaching practice.

Table 4 shows results by country. We find that phone call tutorials are effective in every country, showing scalability and effectiveness across diverse contexts. Results are largest in countries which experienced the longest school closures: Uganda and the Philippines. In both countries, where students were out of school for almost two years and the counterfactual for the control group was very limited schooling, effects are extremely large with the phone call tutorials arm producing improvements of 0.891 standard deviations in Uganda and 0.454 in the Philippines on average ($P < 0.001$ for both). These are also the only countries where there were statistically significant learning gains in the SMS-only arm, with 0.207 gains in Uganda and 0.090 in the Philippines. SMS messages have no statistically significant effect in Nepal and Kenya. These results suggest that SMS messages can work in contexts with the largest need. However, live phone calls may be necessary to strike the balance of being intensive enough to deliver impact that can be sustained and scaled across diverse contexts while remaining cheap and scalable.

Figure 2: Learning Outcomes (SD) at Scale across 5 Countries and Scaled by Government



Notes: This figure shows treatment effects on learning outcomes. Data are presented as treatment effects relative to the control group ± 95 percent confidence intervals. Full statistical results for the treatment effects are presented in Table 1 and 2. Effects are expressed in terms of standard deviations for comparable units. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (for which we report the average level on a scale of 0–4). The colors distinguish between the pooled learning in 5 countries and the Government vs NGO scale-up in Nepal and Philippines.

Table 5 shows learning outcomes across specific proficiencies.²⁵ Results show that the combined phone and SMS arm increases the share of students who get division problems correct by 13.5 percentage points – a 92 percent increase in division (from a control mean of 14.6 percent). These results by proficiency demonstrate that learning outcomes improved substantially in absolute terms, in addition to standardized deviation gains, and across a range of proficiencies. We also find increases in the share of students able to correctly answer place-value problems, and higher-order competencies, such as fractions and word problems. Since fractions were not directly taught during the intervention, this reveals learning extended beyond familiarity with the content taught. These results further reveal dynamic complementarities in skill formation, showing the benefits of learning basic numeracy accrues to learning additional higher-order skills (Cunha and Heckman 2007).

We further explore learning loss and learning recovery along specific proficiencies in Appendix Figure A2. In Uganda, where effects are largest, less than 17 percent of grade 4 students can divide at baseline in the control group. At endline, only 10 percent can divide, showing substantial learning loss. In the treatment group, 48 percent of grade 4 students can divide at endline, fully recovering learning loss due to school closures. Moreover, only 21 percent of grade 5 students in the control group at baseline can divide, revealing that grade 4 students in the treatment group surpass grade 5 students in the status quo. Thus, not only does the intervention facilitate full learning loss recovery, it exceeds typical learning trajectories by nearly an order of magnitude.

In Table A7, we show results on caregivers’ time usage, including on overall educational investments and time spent at work. It is possible that by inducing caregivers to support one child’s education, they invest less in other children in the households’ education. Yet we find evidence of the program crowding in, rather than crowding out, overall time spent on education. The program caused a net increase in the share and frequency of caregivers undertaking educational activities with their children. Moreover, some studies suggest that mothers’ engagement in their child’s education could crowd out their labor market participation (Evans, Jakiela, Knauer 2021). However, we find limited evidence of such a trade-off, potentially since the program is highly efficient, requiring only a short amount of time to produce large learning gains. These results suggests that highly efficient education programs can deliver large learning gains, with minimal risk of crowd-out of other productive educational and work activities.

VI.B Measuring learning by phone

In addition to the main results of the education in emergency programs, we also contribute new evidence on the robustness of remote learning assessment data across five countries. Phone assessment has emerged as a common strategy for large-scale household surveys such as the World Bank Living Standards Measurement Survey (LSMS). A growing literature has started to explore the validity of phone-based assessments to measure learning outcomes (World Bank 2022; Angrist, Bergman, and Matsheng 2022; Gupta et al. 2022; Rodriguez-Segura and Schueler 2022), with emerging evidence suggesting phone assessment can capture meaningful information at high frequency and low cost.

²⁵Table A5 shows additional results for each proficiency of the ASER test.

Table 6 shows the results of five checks we conducted on the validity of our main learning outcomes of learning assessments via phone. Column 1 shows the first robustness check where we compare in-person to phone-based assessment for the exact same students in Kenya. We find no statistically significant difference between these two modes of assessment. An additional test included back-checks, with a random subset of students tested twice on the same competencies. We find a strong relationship as expected, with large positive coefficients and t-statistics ranging from 5 to over 20. We further randomize students to receive different problems of the same proficiency (e.g., four different questions to measure 2-digit addition with carryover). Results show no difference by question, showing accurate estimates of latent ability. Finally, Column 5 shows results from a real-effort question to disentangle effects of the intervention on effort on the test versus cognitive skills. Students were asked to answer several effort tasks, for example figuring out the day of the week or counting zeros and ones.²⁶ We find no statistically significant effects of the interventions on effort, revealing that learning gains are largely a function of cognitive skills.

VI.C Results from Another Education Emergency

COVID-19 disrupted education for over a billion children, causing one of the world’s largest scale education emergencies. However, as Figure 1 shows, many other education emergencies occur beyond COVID-19. Yet little experimental evidence exists in these settings on how to most effectively promote learning.

In addition to the core set of randomized trials assessing impact of phone-based tutorials during COVID-19 school disruptions, an additional education emergency took place during the course of the study – a devastating typhoon in the Philippines which destroyed 4,000 classrooms and disrupted learning for 2 million children. We collected detailed student-level data on who was affected by the typhoon (called “Typhoon Rai”). The typhoon resulted in a shock which further disrupted schooling and learning. Our initial randomization remained unbiased among groups affected by the typhoon, enabling us to assess effectiveness of phone call tutorials in this additional education emergency context.²⁷ The results provide evidence on the disruptive effects of the typhoon on learning and approaches to ensure learning continuity during disruptions.

We start by estimating learning losses due to Typhoon Rai. Results in Table 7 show that the typhoon was associated with reductions in learning by approximately 0.12-0.20 standard deviations (Column 1 and 2). In addition, since randomization of treatments is orthogonal to which students were affected by the typhoon, we can estimate the impact of the phone-based tutorials during this additional emergency. Results in column (4) which includes controls show that despite the

²⁶There is no standard measurement of student effort in the academic literature. However, real-effort tasks range from solving mazes to adding a series of 2-digit numbers. Other proxies of effort include measuring the rate of decline in performance as the test progresses or the effort exerted while filling out an additional survey. Since we aimed to differentiate numerical ability from effort, we chose a problem that required little arithmetic skill, and required non-arithmetic effort.

²⁷There is no statistically significant relationship between randomization to treatment groups and being affected by the typhoon, with p-values of 0.65 and 0.62 for the Phone Call as well as the SMS only treatment, respectively. A full balance table is available on request.

detrimental effects of the typhoon, phone call tutorials continued to be effective, with 0.26 standard deviation learning gains relative to the control group ($P < 0.01$). SMS messages alone were not enough to stem learning losses. These results reveal that the results of effective education in emergencies programs, such as phone call tutorials, can persist across multiple emergency settings.²⁸

VI.D Secondary Outcomes: Parent and Child Beliefs and Non-Cognitive Skills

In addition to learning outcomes, we examine impacts on beliefs about learning. We provide new evidence on children’s perceptions of their own learning in Table 8 Column 1, an outcome rarely explored to date. We find children update their beliefs substantially, by 0.21 standard deviations in the phone call and SMS treatment and by 0.04 in the SMS-only treatment - corresponding to patterns in learning gains observed in Table 1. Column 2 examines caregivers’ beliefs and shows that they observed their child’s learning, updating their beliefs about their child’s level of learning by 0.114 standard deviations in the phone call and SMS treatment and by 0.075 in the SMS only treatment – corresponding broadly to gains observed in Table 1. This reveals parents can learn through noticing (Hanna, Mullainathan, and Schwartzstein 2014) although imperfectly and less accurately than their children. Moreover, caregivers explicitly state that they think their child’s learning has progressed (column 3), with large effects for phone call tutorials, in line with where treatment effects are largest. These results build on a literature exploring parents knowledge of their child’s learning level, enabling them to better support their education (Bergman 2021).

Finally, Table 9 examines impacts on non-cognitive skills. A growing literature highlights the importance of both cognitive as well as non-cognitive skills for future life outcomes, such as graduation from college and labor market outcomes (Jackson 2018). We assess impacts on a series of non-cognitive skills, such as perseverance and ambition, in line with Carlana and La Ferrara (2021). We asked students who had just completed a riddle if they wanted to complete another difficult riddle question (perseverance), and if so, whether they wanted an easier or harder one (ambition). We find sizable effects on these outcomes, with 6.2 percentage point gains in ambition, a 29 percent increase relative to the control mean. We further find positive effects on measures of child well-being, such as enjoying school and worrying less, statistically significant at the 95 percent level and above. There was no effect on these non-cognitive skill outcomes for the SMS-only treatment.

These results reveal that while there is often a debate between whether education should focus on cognitive or non-cognitive skill acquisition, they are not mutually exclusive. Indeed, some educational interventions, such as the phone call tutorials tested in this study, can promote both. It also shows that phone-based tutoring models can impart the types of non-cognitive skills that are often viewed as a benefit of brick-and-mortar schools. This suggests that when emergencies do

²⁸Implementation data suggests that households that were exposed to the typhoon had similar levels of program engagement throughout, and also had high endline survey response rates similar to other studies. For example, typhoon-affected students attended only slightly fewer sessions than non-typhoon affected households (7.39 vs 7.45), and only marginally less time spent per session (20.4 minutes vs. 21 minutes in the final week). This highlights the resilience of the mobile phone program platform to different disaster types, such that even in the midst of extreme weather disruption, households were responding to tutoring and endline calls at high rates, similar to those in other studies and in non-disaster-affected areas.

disrupt education, children can still make progress on both cognitive and non-cognitive skills from targeted phone tutoring programs.

VI.E An External Validity Analysis

Our multi-site randomized trials enable us to uniquely address both internal and external validity challenges simultaneously, enhancing scientific reliability (Maniadis, Tufano, and List 2017).

A prominent alternative approach to addressing external validity concerns includes adjusting for covariates and extrapolating treatment effects from a source setting to a target setting (Hotz, Imbens, and Mortimer 2005; Dehejia, Pop-Eleches, Samii, 2021). We conduct a similar exercise as a comparison, using results from a single setting and rich information on baseline covariates to extrapolate effects across sites. We focus on three core baseline covariates: baseline learning levels — the most predictive covariate of endline learning outcomes — as well as gender and student grade. We first estimate conditional average treatment effects in a source setting controlling for baseline covariates. We then estimate treatment effects in a new target setting based on covariates in that setting using inverse probability weighting. We conduct this exercise for each source and target country. Finally, we compare estimates with randomized trial estimates, a unique validation made possible by this multi-site study.

Table A10 shows our results. While the average treatment effect in India, for example, is around 0.21, predicted treatment effects in India when using Uganda as a source setting are 0.92, very far from the true treatment effect. Predicted effects in India when using source results from Kenya are estimated at 0.12, half of the true treatment effect, and are negative when using Nepal as the source setting. The only source setting that predicts effects in India reliably is the Philippines, estimating an accurate true treatment effect of 0.21. We find similar patterns across the entire matrix of source and target countries, with most predictions far from the true treatment effect. A major factor driving these results is that baseline covariates are fairly similar across sites and we find limited treatment effect heterogeneity. Thus, when extrapolating effects from a given source setting, predicted effects remain largely similar to the source treatment effect, whereas true treatment effects across settings in our randomized trials show more variation. Similar to findings from Dehejia, Pop-Eleches, Samii (2021) these results suggest that “macro” forces — such as the length of school closures in a country or the quality of implementation as shown in Figure A4 — might drive treatment effect heterogeneity more than “micro” forces — such as baseline covariates.

Overall, we find extrapolation based on covariates, a leading approach to address external validity concerns, is not reliably able to predict causal estimates across countries and studies in our setting. This result motivates more multi-site trials such as the ones conducted in this study to reliably assess external validity. A nascent but growing set of studies are using this approach.²⁹

²⁹e.g. Lucas et al. (2014); Banerjee et al. (2015); Banerjee et al. (2017); Bando, Näslund-Hadley, and Gertler (2019), among others.

VI.F Mechanisms

The effectiveness of the phone call tutorials on learning across diverse settings is striking. In this section, we bring data to bear on two mechanisms underpinning the effectiveness of the approach across contexts, as outlined in the conceptual framework: platform and pedagogy.

VI.F.1 Platform: reach at the right level

As described in the conceptual framework, the program’s delivery platform of widely available mobile phones served to enable high take-up and high engagement throughout the program. Data show extremely high engagement in the program, with over 95 percent of treatment households in the phone call arm reached during at least one call.

Detailed monitoring data also shows high week-on-week engagement across sites ranging from 70 to 80 percent, as shown in Figure A1. This highlights that the program reached households using a platform they found easy and convenient to access on a regular basis. This contrasts with the low rates of take-up of other platforms, with well below a third of households typically taking up online, television or radio educational resources. In our study contexts, less than 5 percent of the sample had accessed radio or online education resources in Nepal or the Philippines. A study in Uganda found that only 29 percent of students engaged in radio lessons, 22 percent printed self-study materials, and 12 percent accessed online and TV resources (Uwezo Uganda 2021).

High engagement with mobile phone platforms is also demonstrated by caregivers having very high demand for the phone call tutorials even in low resource and disrupted settings. The control means show that 97 percent of parents state they would like the phone call tutorials, an unusually high level of interest. The phone call tutorials induce even greater interest, bridging the gap to 100 percent interest. Willingness to Pay (WTP) for the program also increases by 4 to 6 percentage points as a result of receiving the calls, revealing the potential to stimulate further demand.³⁰

This high household consent, engagement, and demand is consistent with the notion that phones provide a widely accessible and convenient platform for households to engage with educational content, enabling households to be ‘reached at the right level’.

VI.F.2 Pedagogy: teach at the right level

The program pedagogy – teaching at the right level – was as critical as the platform. Phone call tutorials were designed to target instruction to children’s learning level. In this section, we examine heterogeneous treatment effects, both across and within contexts, to see how much targeted instruction predicts program effectiveness.

We start by exploring conventional heterogeneous treatment effects in Table A6. As column 3 shows, the program is similarly effective for students across the distribution of baseline learning levels. The consistency of impact regardless of students’ starting conditions or characteristics such

³⁰WTP was measured by asking if a math tutoring program were hypothetically offered to the household in the future, how much would they be willing to pay for it. This outcome is constructed as an indicator variable coded 1 if they would be willing to pay.

as baseline learning levels or gender, also shown in Table A6, is consistent with the mechanism of teaching at the right level. Targeted teaching is designed so that instruction meets children where they are regardless of their grade or age or other characteristics. Thus, minimal heterogeneity is likely due to the fact that the program was highly targeted, benefiting each child similarly.³¹

Next we explore heterogeneous treatment effects across contexts. We find that effect variation across trials is consistent with variation in targeted instruction. While results show that phone calls were consistently effective across countries, the magnitude of effects increases in tandem with the order of the trial: Kenya, then Nepal, then India, then Philippines, and finally Uganda. This order of effectiveness tracks the degree of targeted instruction. Detailed monitoring data provides evidence of increasing implementation fidelity as the trials progress. For example, the accuracy of targeted educational instruction increases from a starting point of 50.9 percent of students in Nepal to 81.5 percent on average in Uganda, shown in Appendix Figure A4. These data reinforce the importance of targeted instruction. As the targeting mechanism improved study after study, this coincided with the program becoming ever more effective. Equipped with monitoring data and coordinating mechanisms to learn across trials, these results show that programs have the potential to experience a ‘learning curve’ with improved fidelity (e.g. having ever more targeted instruction) and thus higher impact over time as they are scaled and implemented in new contexts.

In addition to the trial-by-trial improvement in targeting accuracy, we find similar rates of targeting accuracy through both NGO and government delivery. This consistent targeting accuracy across implementer types may explain the similarity in effectiveness of NGO and government teachers. It also adds evidence on the importance of targeting accuracy as a mechanism for impact.

Altogether, the evidence suggests that both a widely accessible mobile platform and effective targeting pedagogy played important roles in program effectiveness. These mechanisms may explain why the approach tested worked across diverse contexts. Consistent with the importance of these mechanisms, a phone education program that did not rely on these two principles had less impact: Crawford et al. (2021) found that phone calls in Sierra Leone that delivered non-targeted lessons based on a mass radio program to students did not improve learning. In this case, it was an accessible platform without targeted pedagogy. To successfully scale across different contexts, ensuring both underlying mechanisms are in place is likely to be key to improve learning.

VI.G Cost-effectiveness

An important feature of the education in emergencies approaches tested in this study is that they are low cost. The primary tool required to implement the program is a mobile phone, a device most households have access to, even in low resource settings (World Bank 2021). Since the approach builds on existing household infrastructure, the main costs are related to content delivery and connecting with families, which are often marginal, such as airtime for phone calls. In addition to

³¹The only margin where we find slight heterogeneity is on parental education. Column 1 in Table A6 suggests the program worked particularly well for students where the caregiver had lower levels of formal education (primary education or less). This suggests that results are strongest when there are fewer alternative education support systems at home. This also reveals that even in low literacy contexts, parents can be effective conduits for quality instruction.

being low cost, the approach has low procurement needs, an attractive feature for governments.

We carefully collected cost data in each trial. Our estimates suggest an average cost per child of the phone call and SMS tutorials of about \$11 per child.³² We benchmark the program’s impacts against other education programs using a variety of approaches. First, we compare raw estimates of effectiveness and cost with similar programs, like tutoring and targeted instruction. Phone call tutorials simulate the benefits of one-on-one tutoring, shown to be one of the most effective educational approaches (Nickow et al. 2020). However, many tutoring programs are high cost. For example, a prominent tutoring program yielded 0.19 to 0.31 SD learning gains at a cost of \$2,500 per child. In low- and middle-income settings, phone calls could provide similar or larger impacts two orders of magnitude more cheaply, enabling scale-up across diverse settings.³³

Second, we use a new cost-effectiveness measure in education that has been estimated for over 150 impact evaluations in low- and middle-income countries called Learning Adjusted Years of Schooling (LAYS), interpreted as a high-quality year of schooling gained (Angrist et al. 2020). We find the program yields 3.9 LAYS per \$100, ranking among the top 10 out of 150 education interventions reviewed.³⁴ This result highlights the potential for phone call tutorials to deliver value to education systems and students in a broad array of contexts, both during and potentially even outside education emergencies. Even before the COVID-19 pandemic, there was high demand for cost-effective education programs to help address the global learning crisis. Our results suggest phone-based targeted tutoring programs, such as the one tested in this study, have the potential to deliver cost-effective learning gains across contexts and with governments.

VI.H A Follow-On Evaluation at Scale

A follow-on scale-up to the five-country randomized controlled trials provides an opportunity to evaluate effects at scale and in a post-pandemic setting. In 2024, the government of Karnataka announced that it had budgeted for a scale-up and planned to roll out the phone call targeted tutorial treatment across 93 “High Needs Blocks” with over 25,000 teachers. This scale is over an order of magnitude larger than the sample in our randomized trials which took place with 500 teachers. Since the scale-up occurred during school term time, teachers were assigned just a few students to tutor, so the workload and time required to tutor remained manageable. We collect data over time for approximately 100,000 students enrolled in the program. Data was collected on

³²This estimate is even cheaper than estimates in the proof-of-concept study in Botswana due to economies of scale as the approach has scaled up, including cost savings such as use of existing pedagogical material, shorter and more efficient training, and streamlined data monitoring systems.

³³The SMS-only treatment, when effective, can be extremely cost-effective, at 41.1 LAYS per \$100. However, while Phone and SMS effects were consistent across all countries, the SMS-only arm impacts varied across contexts, with significant results observed in two contexts with some of the world’s longest school closures: Philippines and Uganda. Given this, it seems plausible that the SMS-only intervention presents a cost-effective option in extreme education emergencies where, for example, calls are not an option or schooling is disrupted for such an extended period of time such that any provision of content is substantially better than the status quo.

³⁴This follows the approach used by the Global Education Evidence Advisory Panel by calculating LAYS across diverse types of education programs. This comparison assesses impact on a set of specific numeracy questions. It is not intended to be a comprehensive examination of all content students would learn in school.

learning levels with a similar learning tool used in the randomized controlled trials, and students were similarly enrolled in grades 3 through 5.

The intervention was conceived as both a complement and substitute to normal schooling. During school terms, the approach could complement schooling to provide extra tutoring and remediation to a subset of students who had yet to acquire basic skills. This would then also position teachers to support learning continuity if a disruption occurred, such as heavy rains during the Monsoon season. Equipped with student phone numbers and the skills and experience needed to provide effective remote instruction, teachers would be able to quickly respond during an education emergency even if schools suddenly closed. Teachers could also provide instruction to a larger share of their class during disruptions with substantial time available due to school closures.

While the government did not randomize rollout of the intervention, we leverage rich data collected on learning levels over time to conduct a difference-in-difference analysis and provide suggestive causal evidence on effectiveness at scale. We estimate the difference-in-difference results as follows. We define the average change in outcomes for each group as:

$$\begin{aligned}\Delta Y_0 &= \bar{Y}_{0,1} - \bar{Y}_{0,0} \quad (\text{Control group: post - pre}) \\ \Delta Y_1 &= \bar{Y}_{1,1} - \bar{Y}_{1,0} \quad (\text{Treated group: post - pre})\end{aligned}$$

The Difference-in-Differences (DiD) estimator is then:

$$\begin{aligned}\hat{\beta}_{\text{DiD}} &= \Delta Y_1 - \Delta Y_0 \\ &= (\bar{Y}_{1,1} - \bar{Y}_{1,0}) - (\bar{Y}_{0,1} - \bar{Y}_{0,0})\end{aligned}$$

Where $\bar{Y}_{d,t}$ is the average outcome for group $d \in \{0, 1\}$ in period $t \in \{0, 1\}$. $d = 1$ is the treated group; $d = 0$ is the control group; $t = 1$ is the post-treatment period; and $t = 0$ is pre-treatment period. In our case, the treatment group is grade 4 students who were exposed to the program between time period 0 and 1 over the course of a school year; the control group are students who were not exposed to treatment, estimated by the baseline difference in learning levels between students in grades 4 and 5. This captures status quo learning in a typical year of schooling for students moving between grades 4 and 5. We assess pre-trends and show learning trajectories progress similarly and slowly across prior years (e.g., the learning trajectory between grades 3 and 4).

Figure A8 shows learning trajectories over time and across treatment and control groups. Results show that while students learn very little in a typical year of school in the status quo, students in the treatment group gain around a full level of numerical competency. In practical terms, students progress roughly from subtraction to multiplication level. Corresponding difference-in-difference estimates are shown in Table A11. Effects are both large in absolute terms as well as highly statistically significant. These results are expressed in terms of average learning levels, rather

than in terms of standard deviations, so while we observe clear positive and large effects, we do not directly compare estimates to results from the multi-country randomized trials. These results suggest effectiveness of the approach at large scale and in an additional post-pandemic setting.

VII Conclusion

In this paper, we present results from large-scale randomized trials evaluating the provision of education in emergency programs across five countries: India, Kenya, Nepal, Philippines, and Uganda. We test multiple scalable models, including government delivery, of remote instruction for primary school children during COVID-19, which disrupted education for over 1 billion schoolchildren worldwide, as well as during an additional emergency, a typhoon, which further disrupted schooling.

Despite heterogeneous contexts, results show that the effectiveness of phone call tutorials can scale across contexts; we find consistently large and robust effect sizes on learning, with average effects of 0.30-0.35 standard deviations. In the subset of trials where we randomized whether the intervention was provided by NGO instructors or government teachers, we find similar effects, indicating effectiveness when delivered within government systems.

These results have relevance to global efforts to support education in emergencies. Emergencies – including conflict, diseases, natural disasters, and climate shocks – routinely shut down schools, affecting millions of students who forget and forgo learning. During these shocks, alternative models are needed to deliver education. The results presented in this paper show that rigorous testing of programs in humanitarian settings is possible; moreover, this study identifies approaches that can scale effectively across contexts and cost-effectively improve learning for students.

Even outside education emergencies, millions of children worldwide learn very little in school, either because they are taught curricula beyond their learning level, or they are unable to access quality instruction since they live in remote areas. Given widespread mobile phone ownership rates globally, phone-based tutoring programs like the one studied here have the potential to maintain schooling continuity and accelerate learning even outside emergency settings or after the initial emergency has eased. The low-cost, high-access, and ease of implementation of phone tutoring could build more resilience into education systems, enabling systems to better withstand frequent shocks, and to more generally utilize cost-effective approaches to address a persistent global learning crisis. In this vein, the government of Karnataka in India is scaling this approach during school and to ready for school disruptions due to Monsoon seasons; our initial quasi-experimental estimates suggest sizable effects at scale.

This study has substantial policy implications. Multiple efforts to scale the intervention studied in this paper are underway around the world. This paper provides critical evidence to inform effectiveness across settings and delivery models. Future research could examine refinements to the delivery model to engage parents further, as well as potential spillovers on student learning in the classroom as teacher practices improve.

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IX Tables

Table 1: Learning Outcomes at Scale Across 5 Countries

	Pooled full sample		Weighted by country-arm	
	(1) Learning (SD)	(2) Learning (SD)	(3) Learning (SD)	(4) Learning (SD)
SMS messages	0.083*** (0.027) [0.003]	0.083*** (0.027) [0.002]	0.138*** (0.028) [0.000]	0.139*** (0.028) [0.000]
Phone call and SMS	0.327*** (0.025) [0.000]	0.321*** (0.025) [0.000]	0.414*** (0.026) [0.000]	0.408*** (0.026) [0.000]
Observations	8902	8902	8902	8902
Control Mean	1.764	1.764	1.764	1.764
P-Val: SMS vs Phone Call + SMS	0.000	0.000	0.000	0.000
Country Fixed Effects	Yes	Yes	Yes	Yes
Grade Fixed Effects	No	Yes	No	Yes
Countries Included	All 5	All 5	All 5	All 5

Notes: This table reports treatment effects on learning outcomes for a pooled sample as well as samples weighted by country-arm. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). “Pooled full sample” weighs the observations across all the countries equally. “Weighted by country-arm” accounts for the differences in sample sizes among countries. Column (1) measures the learning differences between the treatment and control group, measured in standard deviations, with country fixed effects only. Column (2) includes baseline controls and country and grade fixed effects. Column (3) and column (4) weigh the results by country-arm. Standard errors are in parentheses and p-values are in square brackets. Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct, and 4 if students got division correct. For example, in this table, average student levels in the control group at endline were 1.764, which is the equivalent of a student knowing between addition and subtraction level.

Table 2: Learning Outcomes for Government Scale Models

	All Government Arms	Nepal	Philippines
	(1)	(2)	(3)
	Learning (SD)	Learning (SD)	Learning (SD)
SMS messages	0.074* (0.040) [0.064]	0.056 (0.061) [0.363]	0.090* (0.048) [0.060]
Phone call and SMS - NGO	0.263*** (0.046) [0.000]	0.111* (0.067) [0.096]	0.434*** (0.057) [0.000]
Phone call and SMS - Gov	0.314*** (0.050) [0.000]	0.170** (0.067) [0.011]	0.482*** (0.070) [0.000]
Observations	4941	2625	2316
Control Mean	1.589	1.718	1.473
P-Val: NGO vs Gov	0.333	0.381	0.534
Country Fixed Effects	Yes	No	No
Countries Included	Nepal, Philippines	Nepal	Philippines

Notes: This table reports learning outcomes comparing implementation delivery models (NGO and Government). Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). Column (1) pools both countries where we tested implementation scale up in government systems. Column (2) shows the results for Nepal alone, and column (3) shows results for Philippines alone. Standard errors are in parentheses and p-values are in square brackets. Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct, and 4 if students got division correct.

Table 3: Spillovers to the System: Changing Teacher Beliefs and Practices

	Teacher practices		Teacher beliefs	
	(1) Get parents involved	(2) Targets to student level	(3) Would be teacher again	(4) Helped students' learning
Teacher implemented call	0.088* (0.050) [0.076]	0.093** (0.046) [0.043]	0.158*** (0.045) [0.000]	0.106* (0.058) [0.067]
Observations	290	290	290	290
Control mean	0.721	0.769	0.735	0.408

Notes: This table shows treatment effects on teacher practices and beliefs. This includes government teachers in Nepal who were randomly assigned to implement the phone call program. We successfully spoke to 83 percent of the government mathematics teachers who were in the schools identified by local governments. Of the eligible grade 3-5 math teachers we spoke to, 81 percent expressed interest in participating in the study. Of this eligible pool of 301 government teachers, 50 percent were randomly selected to implement. The endline response rate was 96 percent and was balanced across the treatment and control groups. Column (1) shows effects on teachers' beliefs that they can get parents involved in their child's education. Column (2) shows effects on teachers saying they tailor student feedback based on students' individual understanding and skill level. Column (3) shows effects on teachers saying if they had to choose their profession again, they would still be a teacher. Column (4) shows effects on teachers saying they believe they helped their students with their maths skills. These are all dummy variables coded as 1 if respondents strongly agreed on a likert scale. Standard errors are in parentheses and p-values are in square brackets.

Table 4: Learning Outcomes by Country

	India	Kenya	Nepal	Philippines	Uganda
	(1)	(2)	(3)	(4)	(5)
	Learning (SD)	Learning (SD)	Learning (SD)	Learning (SD)	Learning (SD)
SMS messages		-0.020 (0.039) [0.606]	0.049 (0.065) [0.450]	0.090* (0.048) [0.060]	0.207*** (0.057) [0.000]
Phone call and SMS	0.212*** (0.067) [0.002]	0.085** (0.038) [0.025]	0.140** (0.061) [0.023]	0.454*** (0.050) [0.000]	0.891*** (0.054) [0.000]
Observations	668	1985	2625	2316	1308
Control Mean	2.799	1.420	1.718	1.473	1.803
P-Val: SMS vs Phone Call + SMS		0.004	0.096	0.000	0.000

Notes: This table reports learning outcomes by country. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). Standard errors are in parentheses and p-values are in square brackets. Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table 5: Learning Outcomes by Multiple Proficiencies

	Division	Innumerate	Other Proficiencies		
	(1) Division	(2) No operations	(3) Place value	(4) Word problems	(5) Fractions
SMS messages	0.029*** (0.010) [0.006]	0.001 (0.014) [0.929]	0.007 (0.015) [0.646]	0.045*** (0.015) [0.003]	0.044*** (0.012) [0.000]
Phone call and SMS	0.135*** (0.011) [0.000]	-0.042*** (0.012) [0.000]	0.047*** (0.014) [0.000]	0.071*** (0.013) [0.000]	0.064*** (0.011) [0.000]
Observations	7163	7163	7163	7163	7163
Control Mean	0.146	0.230	0.639	0.634	0.209
P-Val: SMS vs Phone Call + SMS	0.000	0.000	0.003	0.048	0.070
Country Fixed Effects	Yes	Yes	Yes	Yes	Yes
Grade Fixed Effects	Yes	Yes	No	No	No
Countries Included	4	4	4	4	4

Notes: This table reports learning outcomes across different proficiencies. Column (1) highlights the learning gains in terms of share of students who learned division. Column (2) shows the share of students who could not correctly answer any of the basic numeracy operations (referred to as innumerate). Columns (3), (4), and (5) show the share of students who could correctly answer place-value problems, word problems, and fractions, respectively. The place-value and higher-order questions were not asked of the Grade 1 and 2 students in Kenya as these were not covered in their standard school curriculum at these ages. Standard errors are in parentheses and p-values are in square brackets.

Table 6: Robustness Checks on Learning Outcomes

	Phone vs In Person	Backchecks		Random Problem	Effort
	(1) Learning (SD)	(2) Add Q2	(3) Divide Q2	(4) Learning (SD)	(5) Effort Task
Assessment Mode	0.063 (0.054) [0.244]				
Add		0.473*** (0.085) [0.000]			
Divide			0.653*** (0.029) [0.000]		
Random Order 2				0.064 (0.067) [0.338]	
Random Order 3				0.023 (0.067) [0.734]	
Random Order 4				-0.035 (0.066) [0.596]	
SMS messages					-0.004 (0.019) [0.821]
Phone call and SMS					-0.021 (0.019) [0.257]
Observations	1985	708	708	2617	5048
Control Mean	1.420	0.930	0.309	1.718	0.435
Countries Included	Kenya	India	India	Nepal	3 Countries

Notes: This table reports the robustness checks on learning outcome measurement. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). Column (1) shows the results on average student level for assessments via phone relative to the same assessment in-person. Column (2) and (3) show backcheck results on addition and division. Column (4) shows results of randomizing students to receive different problems of the same proficiency. Column (5) shows results from effort questions. Standard errors are in parentheses and p-values are in square brackets. In column 1, given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table 7: Learning During Other Education in Emergencies - Typhoon in the Philippines

	Typhoon Effects			
	(1) Learning (SD)	(2) Learning (SD)	(3) Learning (SD)	(4) Learning (SD)
Affected by Typhoon	-0.198*** (0.052) [0.000]	-0.116** (0.050) [0.021]		
Control × Typhoon Effect			-0.228** (0.114) [0.046]	-0.108 (0.109) [0.321]
SMS messages × No Typhoon Effect			0.071 (0.074) [0.335]	0.073 (0.070) [0.299]
SMS messages × Typhoon Effect			-0.079 (0.086) [0.359]	0.008 (0.082) [0.923]
Phone call and SMS × No Typhoon Effect			0.433*** (0.080) [0.000]	0.439*** (0.076) [0.000]
Phone call and SMS × Typhoon Effect			0.203** (0.102) [0.047]	0.259*** (0.098) [0.008]
Observations	1647	1647	1647	1647
Countries Included	Philippines	Philippines	Philippines	Philippines
Controls	None	Ed and Bsl Level	None	Ed and Bsl Level

Notes: This table reports the effect of a typhoon on learning outcomes in the Philippines. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). Column (1) shows the effect on learning of students affected by the typhoon. Column (2) accounts for the baseline distribution and the caregiver’s education. Column (3) and (4) show the results of the SMS messages alone as well as phone tutorial treatments with and without controls. Standard errors are in parentheses and p-values are in square brackets. Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table 8: Parent and Child Beliefs

	Beliefs about child's level and progress			Caregiver's program interest	
	(1) Child estimates level	(2) Caregiver estimates level	(3) Child progressed	(4) Wants program	(5) WTP
SMS messages	0.040* (0.024) [0.097]	0.075*** (0.024) [0.002]	0.010 (0.015) [0.522]	-0.019 (0.012) [0.127]	0.060*** (0.020) [0.003]
Phone call and SMS	0.210*** (0.023) [0.000]	0.114*** (0.023) [0.000]	0.104*** (0.014) [0.000]	0.034*** (0.011) [0.001]	0.045** (0.020) [0.025]
Observations	8798	9188	6188	8918	3777
Control mean	2.347	2.285	0.350	0.968	0.434
P-Val: SMS vs Phone Call + SMS	0.000	0.079	0.000	0.000	0.452
Country Fixed Effects	Yes	Yes	Yes	Yes	Yes
Grade Fixed Effects	Yes	Yes	Yes	Yes	Yes
Countries Included	All	All	4	All	Philippines & Uganda

Notes: This table shows treatment effects on secondary outcomes related to parental and child beliefs, and demand for the program. Column (1) measures the child's belief about their level of learning, from beginner (0) to division (4), in standard deviations. Column (2) measures parents beliefs about their child's level on the same scale. Column (3) shows treatment effects on an indicator variable for caregivers being very confident that their child has progressed in their learning. Column (4) shows effects on a dummy variable that the caregiver would like access to a phone-based maths tutoring program in the future, and Column (5) shows treatment effects on a dummy variable that caregivers would be willing to pay money to access a phone-based maths tutoring program in the future. Standard errors are in parentheses and p-values are in square brackets.

Table 9: Child Non-Cognitive Outcomes

	Non-cognitive Skills		Wellbeing	
	(1) Perseverance	(2) Ambition	(3) Enjoys school	(4) Often worried
SMS messages	0.010 (0.010) [0.308]	0.008 (0.011) [0.465]	0.011 (0.011) [0.310]	-0.001 (0.011) [0.957]
Phone call and SMS	0.029*** (0.009) [0.001]	0.062*** (0.011) [0.000]	0.023** (0.009) [0.014]	-0.020** (0.010) [0.043]
Observations	8962	8962	8880	8121
Control mean	0.833	0.211	0.821	0.244
P-Val: SMS vs Phone Call + SMS	0.042	0.000	0.201	0.060
Country Fixed Effects	Yes	Yes	Yes	Yes
Grade Fixed Effects	Yes	Yes	Yes	Yes
Countries Included	All	All	All	All

Notes: This table shows treatment effects on children’s non-cognitive outcomes, including perseverance and wellbeing. Column (1) measures impacts on a dummy variable for whether the child wanted to answer a second riddle question after answering an initial one, following the Carlana and La Ferrara (2021) measure of perseverance. Column (2) measures impacts on an indicator variable for whether the student wanted to answer a second more difficult riddle question, indicating ambition. This was coded 0 if they didn’t want to answer a second riddle question or they wanted an easier question. Column (3) shows effects on a dummy variable for whether the student says they enjoy school very much. Column (4) shows effects on an indicator for whether the child has many worries or is often worried, from the children’s “Strengths and Difficulties Questionnaire”. Standard errors are in parentheses and p-values are in square brackets.

A Appendix Tables

Table A1: Representativeness of sample at endline

	Baseline Variables				
	(1) Baseline Learning	(2) Grade 3	(3) Grade 4	(4) Grade 5	(5) Student is Female
Randomized to receive endline survey	-0.009 (0.021) [0.679]	-0.000 (0.006) [0.967]	-0.003 (0.006) [0.558]	0.004 (0.005) [0.462]	-0.005 (0.010) [0.634]
Observations	12707	16936	16936	16936	16936
Comparison Mean	1.39	0.35	0.37	0.28	0.51
Country Fixed Effects	Yes	Yes	Yes	Yes	Yes
Countries Included	All 5	All 5	All 5	All 5	All 5

Notes: This table reports balance on baseline characteristics of the sample that was randomly selected to be interviewed at endline, compared to the complete study sample. Column (1) shows the representativeness by baseline learning, column (2) (3) and (4) by student grade in the eligible sample, and column (5) by student sex. These are all of the variables collected across all 5 countries at baseline. On baseline learning, note that in Nepal, a randomly selected half of the sample received a full baseline survey that included learning level. Learning is expressed in standardized units (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). In Kenya, school assessment grades were used in lieu of a baseline levelling assessment, and not all Kenyan students had pre-existing student grades. Sample means for grades are calculated for students in countries with the relevant grades. Standard errors are in parentheses and p-values are in square brackets. In column 1, given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include average student learning level in the control group at baseline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table A2: Reach and Attrition

	Households Reached		Correlation in Attempts
	(1) Reached at Endline	(2) Reached at Endline	(3) Baseline Learning
SMS messages	-0.001 (0.010) [0.913]	-0.001 (0.010) [0.912]	
Phone call and SMS	0.008 (0.010) [0.445]	0.008 (0.010) [0.430]	
Number of attempts to reach HH			0.001 (0.005) [0.889]
Observations	12331	12331	7187
Control mean	0.799	0.799	
Country Fixed Effects	Yes	Yes	Yes
Grade Fixed Effects	No	Yes	Yes
Countries Included	All 5	All 5	All 5

This table reports the differences in the households reached. Column (1) measures the households reached at endline. Column (2) measures the same but accounts for the students' grade level. Column (3) shows the effect of the number of attempts to reach a household to the baseline learning of the student. The control means show that on average 80 percent of households were reached at endline. Standard errors are in parentheses and p-values are in square brackets.

Table A3: Balance on baseline characteristics

	Baseline Variables				
	(1) Baseline Learning	(2) Grade 3	(3) Grade 4	(4) Grade 5	(5) Female
SMS messages	0.035 (0.027) [0.191]	-0.001 (0.013) [0.955]	0.007 (0.013) [0.571]	-0.006 (0.008) [0.393]	0.016 (0.012) [0.181]
Phone call and SMS	0.007 (0.029) [0.801]	-0.001 (0.012) [0.906]	0.008 (0.012) [0.520]	-0.006 (0.007) [0.352]	0.004 (0.011) [0.725]
Observations	9376	12331	12331	12331	12331
Control Mean	1.47	0.34	0.34	0.24	0.53
P-Val: SMS vs Phone Call + SMS	0.36	0.95	0.96	0.98	0.29
Country Fixed Effects	Yes	Yes	Yes	Yes	Yes
Countries Included	All 5	All 5	All 5	All 5	All 5

This table reports the balance across treatment groups at baseline for the sample randomly selected to be interviewed at endline. Column (1) shows balance on baseline learning, column (2) (3) and (4) by student grade in the eligible sample, and column (5) by student sex. These are all of the variables collected across all 5 countries at baseline. On baseline learning, note that in Nepal, a randomly selected half of the sample received a full baseline survey that included learning level. Learning is expressed in standardized units (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). In Kenya, school assessment grades were used in lieu of a baseline levelling assessment, and not all Kenyan students had pre-existing student grades. Control means for grades are calculated for students in all countries with relevant grades. Standard errors are in parentheses and p-values are in square brackets. In column 1, given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include average student learning level in the control group at baseline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table A4: Balance on baseline characteristics: Nepal teacher sample

	(1)	(2)	(3)	(4)	(5)	(6)
	Female	Years teaching	High school	Bachelors	Masters	Speaks English
Treatment Teacher	0.037 (0.056) [0.504]	0.567 (1.167) [0.627]	0.004 (0.057) [0.947]	0.004 (0.056) [0.941]	-0.034 (0.039) [0.380]	0.044 (0.056) [0.434]
Observations	301	301	301	301	301	301
Control mean	0.353	13.387	0.427	0.373	0.147	0.360

This table reports the balance across Nepal's teacher treatment and control group at baseline. Column (1) shows balance on teacher gender. Column (2) shows balance based on teacher's years of experience. Columns (3) to (5) show balance on the share of teachers whose highest level of education completed is high school, bachelors degree, or masters degree, respectively. Column (6) shows balance on the share of teachers who speak English. Standard errors are in parentheses and p-values are in square brackets.

Table A5: Learning Outcomes by All 4 Operations Taught

	(1)	(2)	(3)	(4)
	Addition correct	Subtraction correct	Multiplication correct	Division correct
SMS messages	-0.005 (0.012) [0.675]	0.039*** (0.013) [0.003]	0.067*** (0.015) [0.000]	0.030*** (0.010) [0.004]
Phone call and SMS	0.041*** (0.010) [0.000]	0.119*** (0.012) [0.000]	0.193*** (0.014) [0.000]	0.138*** (0.011) [0.000]
Observations	9148	9148	7163	7163
Control Mean	0.775	0.586	0.363	0.152
P-Val: SMS vs Phone Call + SMS	0.000	0.000	0.000	0.000
Country Fixed Effects	Yes	Yes	Yes	Yes
Grade Fixed Effects	Yes	Yes	Yes	Yes
Countries Included	5	5	4	4

Notes: This table reports learning outcomes across the 4 different proficiencies taught in the curriculum and comprising the standardized learning outcome. Column (1) highlights the learning gains in terms of share of students who got addition correct. Columns (2), (3), and (4) show the share of students who could correctly answer subtraction, multiplication, and division respectively. The higher-order questions were not asked of the Grade 1 and 2 students in Kenya as these were not covered in their standard school curriculum at these ages. Standard errors are in parentheses and p-values are in square brackets.

Table A6: Heterogeneity

	Caregiver Education	Gender	Baseline Level
	(1)	(2)	(3)
	Learning (SD)	Learning (SD)	Learning (SD)
SMS messages	-0.042 (0.111) [0.703]	0.110*** (0.039) [0.005]	0.064** (0.028) [0.025]
Phone call and SMS	0.206** (0.105) [0.050]	0.333*** (0.035) [0.000]	0.360*** (0.027) [0.000]
Primary	-0.010 (0.083) [0.909]		
Secondary Plus	0.304*** (0.078) [0.000]		
SMS messages × Primary	0.097 (0.127) [0.448]		
SMS messages × Secondary Plus	0.133 (0.116) [0.253]		
Phone call and SMS × Primary	0.215* (0.121) [0.075]		
Phone call and SMS × Secondary Plus	0.120 (0.110) [0.273]		
Female=1		0.063* (0.036) [0.085]	
SMS messages × Female=1		-0.053 (0.053) [0.314]	
Phone call and SMS × Female=1		-0.014 (0.048) [0.773]	
Baseline Level			0.234*** (0.021) [0.000]
SMS messages × Baseline Level			0.037 (0.030) [0.204]
Phone call and SMS × Baseline Level			0.040 (0.027) [0.136]
Observations	7794	8902	7035
Control Mean	1.764	1.764	1.764
P-Val: SMS vs Phone Call + SMS	0.030	0.000	0.000
Countries Included	All 5	All 5	All 5
Country Fixed Effects	Yes	Yes	Yes

Notes: This table reports heterogeneous treatment effects on learning outcomes. Learning refers to how a child scores on four basic numeracy options: no operations correct, addition, subtraction, multiplication, and division (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). Column (1) shows the results between having caregivers who reached primary education or less and caregivers who reached secondary education or more. Column (2) shows the results between males and females. Column (3) shows the results across baseline learning levels. Standard errors are in parentheses and p-values are in square brackets. Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table A7: Potential for Crowdout

	Caregiver supported child educ		Caregiver unemployment	
	(1) Did educ activities	(2) Often did educ activities	(3) Unemployed	(4) Mother unemployed
SMS messages	0.044*** (0.009) [0.000]	0.032** (0.013) [0.013]	-0.019 (0.021) [0.358]	-0.029 (0.026) [0.271]
Phone call and SMS	0.042*** (0.008) [0.000]	0.029** (0.012) [0.015]	0.012 (0.019) [0.513]	0.019 (0.024) [0.415]
Observations	8899	8899	3681	2446
Control mean	0.865	0.494	0.429	0.429
P-Val: SMS vs Phone Call + SMS	0.785	0.784	0.119	0.059
Country Fixed Effects	Yes	Yes	Yes	Yes
Grade Fixed Effects	Yes	Yes	Yes	Yes
Countries Included	All	All	Philippines & Uganda	Philippines & Uganda

Notes: This table shows treatment effects on outcomes that could indicate the program crowds-out or crowds-in other education and labour market activities. Column (1) measures impacts on a dummy variable for whether the caregiver says they did any educational activities with their child over the past 3 weeks. Column (2) shows how often the parent did any educational activities with their child 3 or more times in the past week. Column (3) shows effects on caregiver unemployment, and column (4) restricts this unemployment estimate to households reporting that mothers are the primary educational caregiver in the household. Standard errors are in parentheses and p-values are in square brackets.

Table A8: Trial Description

	India	Kenya	Nepal	Philippines	Uganda
Sample size	850	6724	3732	3492	2138
Randomly selected endline sample size	765	3556	3351	3164	1495
Student grades	3-5	1-2	3-5	3-4	3-5
NGO Delivery	✓	✓	✓	✓	✓
Government Delivery			✓	✓	
Unit of randomization	Household	Cluster - school grade	Household	NGO: Household; Gov: Cluster - school grade:	Household
Stratification variables used in block randomization	Student baseline level, school, female	School	Local Government, Parent perception of student baseline level	NGO: Region, Student baseline level; Gov: school, grade, Student baseline level	Student baseline level, Attends government school, previous education program participation
Included Combined Phone Call and SMS treatment	✓	✓	✓	✓	✓
Included SMS only treatment	✗	✓	✓	✓	✓
Implementer type	NGO (Alokit)	NGO public-private partnership (NewGlobe)	Government (Ministry of Education), World Bank, NGOs (Teach for Nepal, Street Child)	Government (Department of Education), Research NGO (Innovations for Poverty)	NGO (Building Tomorrow)
Number of weeks of implementation	8	12	16	8	8
Dates	Apr 21-Jul 21	Dec 20-Apr 21	Jan 21-Jul 21	Aug 21-Jul 22	Oct 21-Jan 22
Administrative units in study	1 state	30 counties	All 7 provinces	3 regions	9 districts

This table summarizes the key features of the intervention across countries. Implementers of the interventions included government education ministries and NGOs.

Table A9: Learning Outcomes at Scale Across 5 Countries, including MHT-adjusted p-values

	Pooled full sample		Weighted by country-arm	
	(1) Learning (SD)	(2) Learning (SD)	(3) Learning (SD)	(4) Learning (SD)
SMS messages	0.083*** (0.027) [0.003] {0.003}	0.083*** (0.027) [0.002] {0.002}	0.138*** (0.028) [0.000] {0.001}	0.139*** (0.028) [0.000] {0.001}
Phone call and SMS	0.327*** (0.025) [0.000] {0.001}	0.321*** (0.025) [0.000] {0.001}	0.414*** (0.026) [0.000] {0.001}	0.408*** (0.026) [0.000] {0.001}
Observations	8902	8902	8902	8902
Control Mean	1.764	1.764	1.764	1.764
P-Val: SMS vs Phone Call + SMS	0.000	0.000	0.000	0.000
Country Fixed Effects	Yes	Yes	Yes	Yes
Grade Fixed Effects	No	Yes	No	Yes
Countries Included	All 5	All 5	All 5	All 5

Notes: This table reports treatment effects on learning outcomes for a pooled sample as well as samples weighted by country-arm. Learning is expressed in standardized units (measured on a scale of 0–4 and expressed in terms of standard deviations, standardized using the control group mean and standard deviation at endline). This table presents the same results as Table 1 but also includes False Discovery Rate (FDR) sharpened q-values to adjust for multiple hypothesis testing. Standard errors are in parentheses and p-values are in square brackets. Curly brackets show the sharpened q-values (Anderson, 2008). Given learning outcomes are standardized to have a control mean of zero, we include in the control mean row a more intuitive comparison. We include unstandardized average student learning level in the control group at endline. This variable takes the value of 0 for students who got zero operations correct, 1 if students got addition correct, 2 if subtraction was correct, 3 if multiplication was correct and 4 if students got division correct.

Table A10: The Limits of Extrapolating Treatment Effects Across Countries Based on Covariates

Source Country for Covariate-Based Prediction					
Target Country	Uganda	Kenya	India	Philippines	Nepal
India	0.921	0.121	0.218	0.217	-0.071
Kenya	0.820	0.061	0.256	0.749	0.265
Nepal	0.892	0.051	0.215	0.276	0.106
Philippines	0.877	0.048	0.224	0.372	0.148
Uganda	0.893	0.085	0.223	0.331	0.052
Total	0.881	0.073	0.227	0.389	0.100

Notes: This table extrapolates treatment effects based on covariates in a given source country to a new target country using inverse probability weighting based on key baseline covariates: baseline learning level, gender, and student grade.

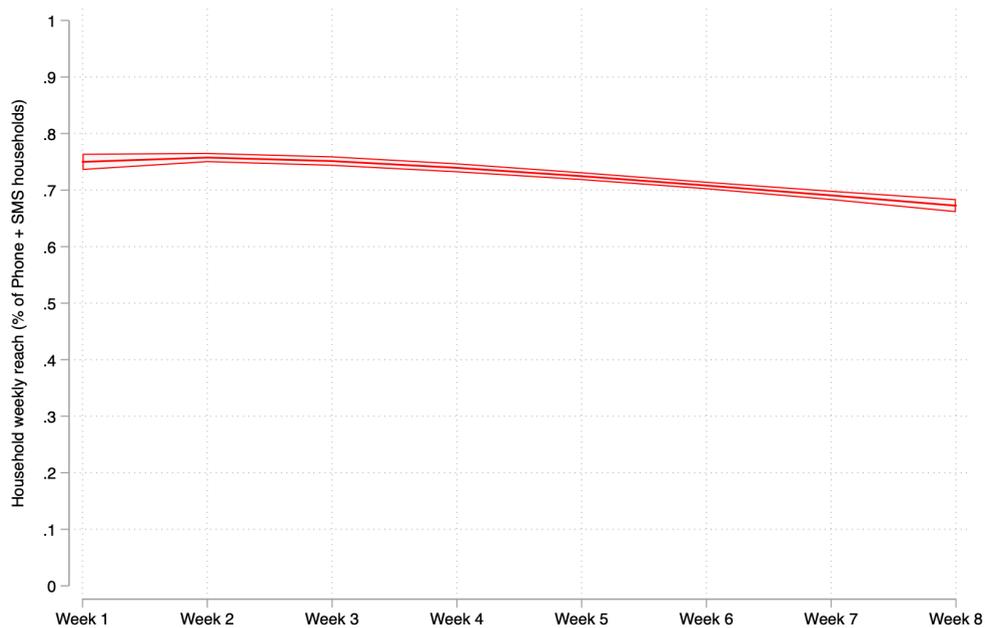
Table A11: Difference-in-Difference Average Learning Level Gains at Scale in India

	Treatment Difference	Control Difference	Difference-in-Difference Estimate
Coefficient	1.25	0.14	1.11
Standard Error	(0.01)	(0.01)	(0.02)
P-Value			< 0.001
N			96,953

Notes: This table shows the difference-in-difference estimation. We show average learning levels for grade 4 students exposed to treatment as well as average learning trajectories in the control group status quo estimated by baseline differences in learning levels between grades 4 and 5. These learning levels are based on data for over 25,000 teachers and using data from around 100,000 children. Average learning levels are coded as follows: 1 = addition; 2 = subtraction; 3 = multiplication; 4 = division. A learning gain of 1 level corresponds to an operational level gained, such as moving from subtraction to multiplication.

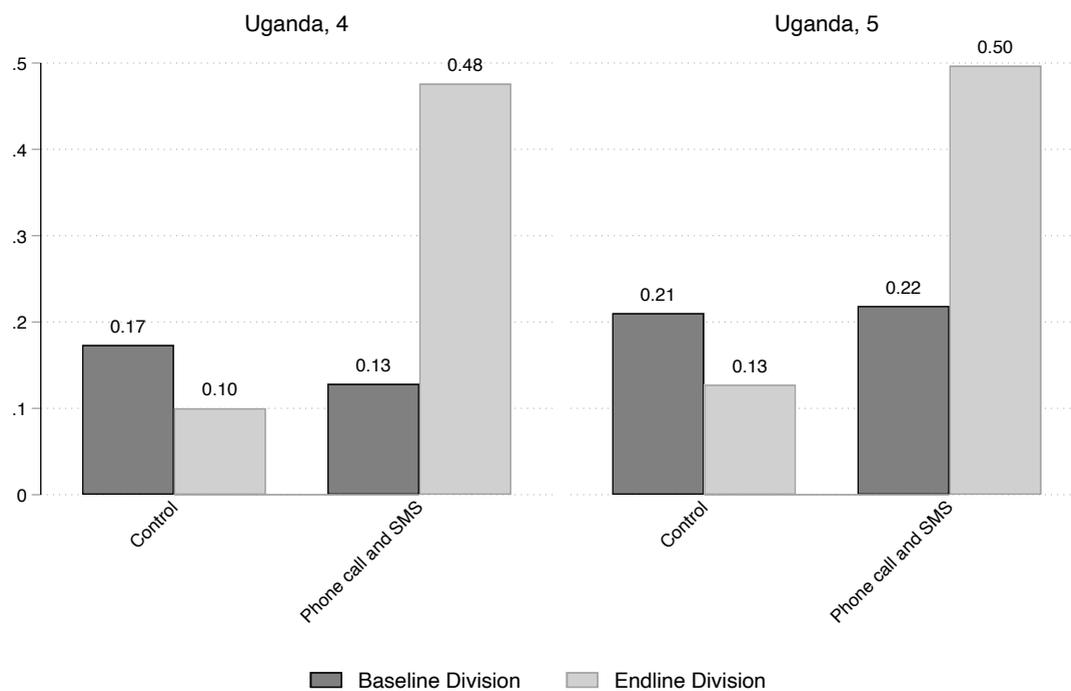
B Appendix Figures

Figure A1: Weekly Household Engagement in Phone Calls



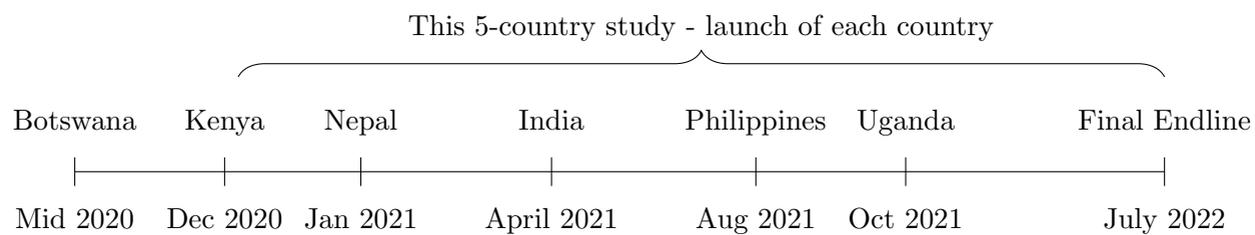
Notes: This figure shows average weekly engagement of households in the phone call and SMS arm. Each week, engagement is coded as 1 if the household answered the call and engaged. This includes data from Nepal, Uganda and the Philippines where we have detailed and easily comparable weekly monitoring data.

Figure A2: Learning Losses & Gains - Can Divide, Uganda, by Grade



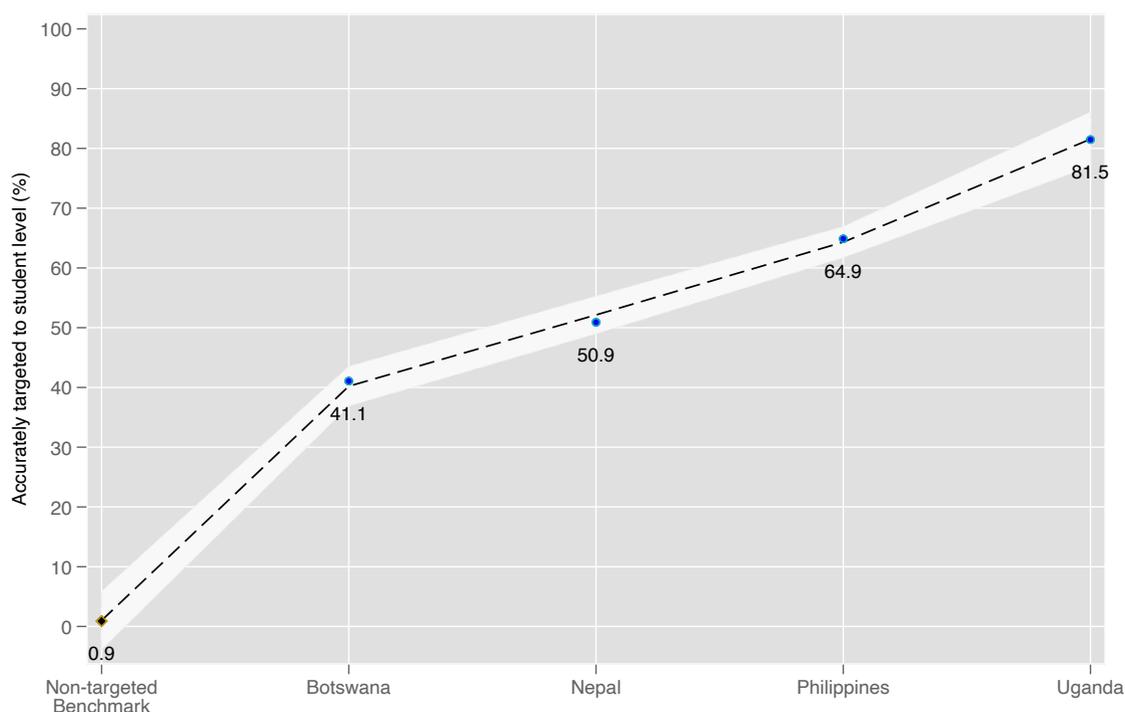
Notes: This figure shows the proportion of Grade 4 and Grade 5 students in Uganda that know how to correctly perform division. The shades distinguish the baseline and endline proportion. In each panel, the two bars on the left show the outcomes for students who did not receive the intervention; the two on the right show the outcomes for those who received the SMS exercises and phone call tutoring. The figure shows that there is significant learning loss for control students between baseline and endline. It also shows that there is only a modest increase in the year-on-year share of students at division-level proficiency between grades 4 and 5 in the status quo. Finally, it shows the share of students able to perform division in the treatment group both recovers learning loss and far exceeds the year-on-year progress in division proficiency, with treated students overtaking the subsequent grade's proficiency level at baseline.

Figure A3: Timeline of trials



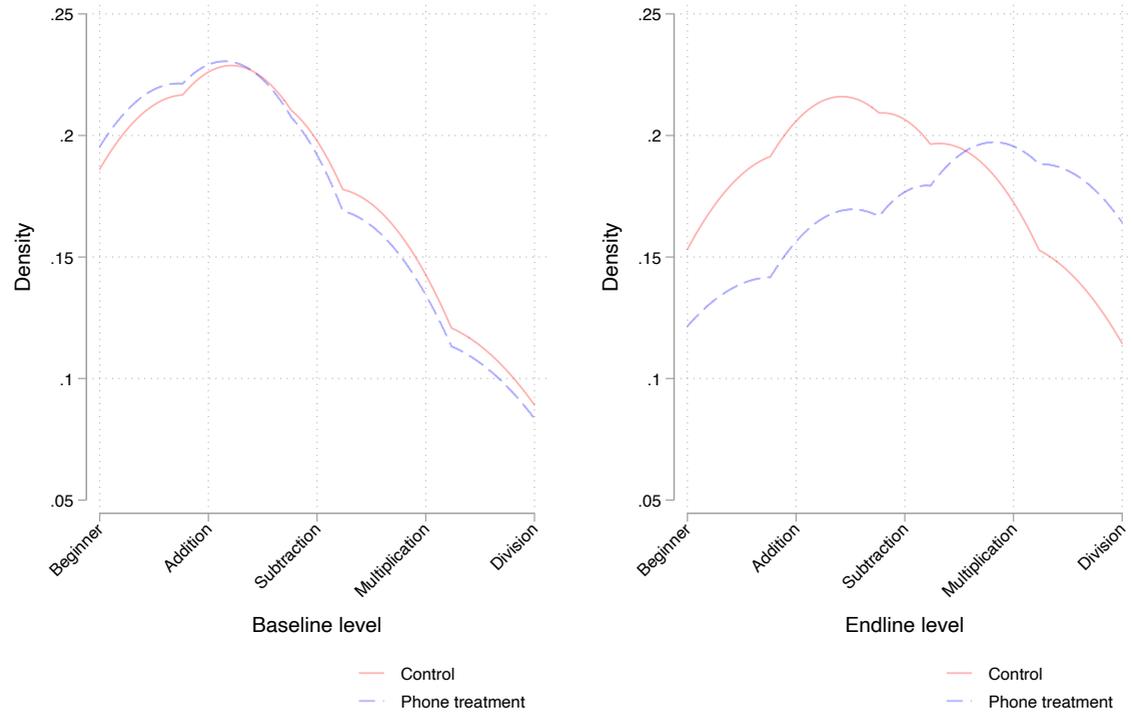
Notes: This figure shows the timing of the implementation across studies. The proof of concept study was conducted in Botswana in 2020. It was followed by five studies across five countries. An endline assessment was conducted a few months after each implementation period ended; the endline for the latest replication was in July 2022.

Figure A4: Learning Curve – Improved Implementation and Targeted Instruction Across Trials



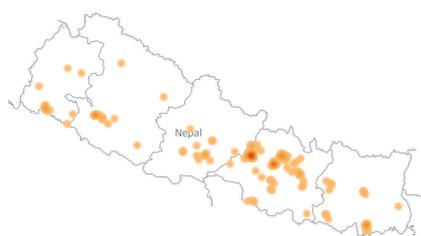
Notes: This figure is adapted from Angrist et al. (2023) in the *AEA Papers and Proceedings* which uses monitoring data to estimate how targeted instruction was across studies where enough monitoring data existed. Targeted instruction is defined as whether instructors taught students at their level each week, and is then averaged across all eight weeks. For example, if a child did not know addition and they were taught addition, the instruction was well targeted; if a child did know addition but was still taught it rather than moving on to subtraction, instruction was not well targeted. Estimates are also included from the Botswana proof-of-concept study (Angrist, Bergman, and Matsheng 2022). Benchmark estimates from control groups of teaching at the right level studies (Banerjee et al. 2017) are also included to show how often instruction is targeted in the status quo. These estimates track learning outcome progress, with learning improving from Botswana (0.12 standard deviations when instruction was targeted 41.1 percent of the time) to Nepal (0.14 standard deviations when instruction was targeted 50.9 percent of the time) to the Philippines (0.44 standard deviations when instruction was targeted 64.9 percent of the time) to Uganda (0.88 standard deviations when instruction was targeted 81.5 percent of the time).

Figure A5: Distribution of learning at baseline and endline across groups

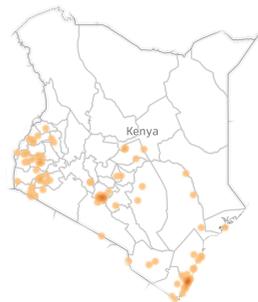


Notes: These figures show the distribution of student math levels by control group and phone call treatment group at baseline (left) and endline (right). ‘Level’ is the highest level students achieved in the ASER-based assessment which starts at Beginner (can’t perform any numerical operations), and then proceeds to Addition, Subtraction, Multiplication, and ends at Division. Additional proficiencies were also tested, such as word problems and fractions, but the core competencies listed above comprise the common ASER learning module across countries. These density plots include all countries except Kenya, since in Kenya baseline school exam scores were used instead of a new baseline assessment.

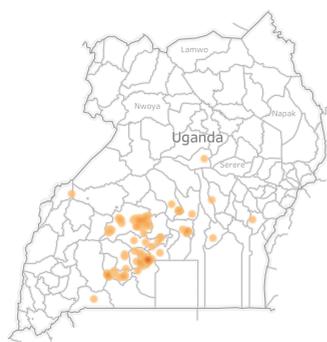
Figure A6: Maps of the Study Sample by Country



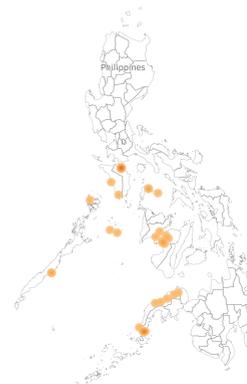
(a) Nepal



(b) Kenya



(c) Uganda

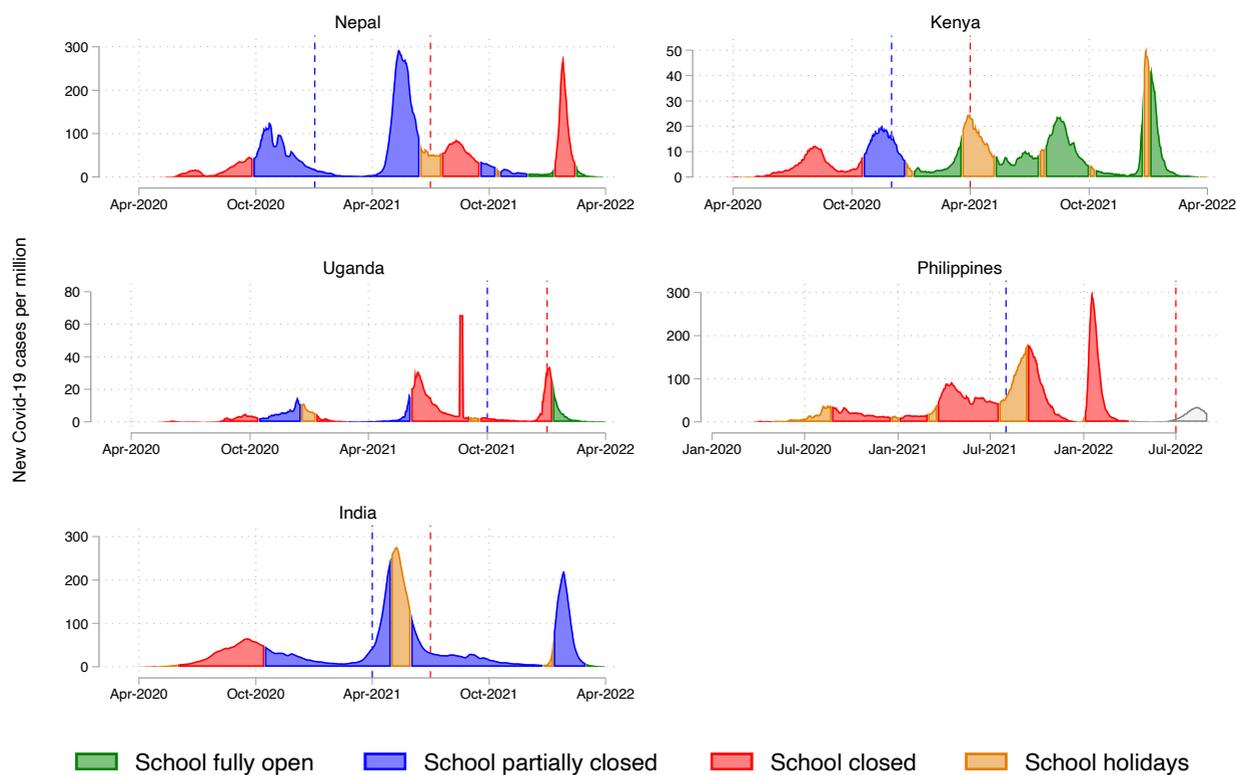


(d) Philippines



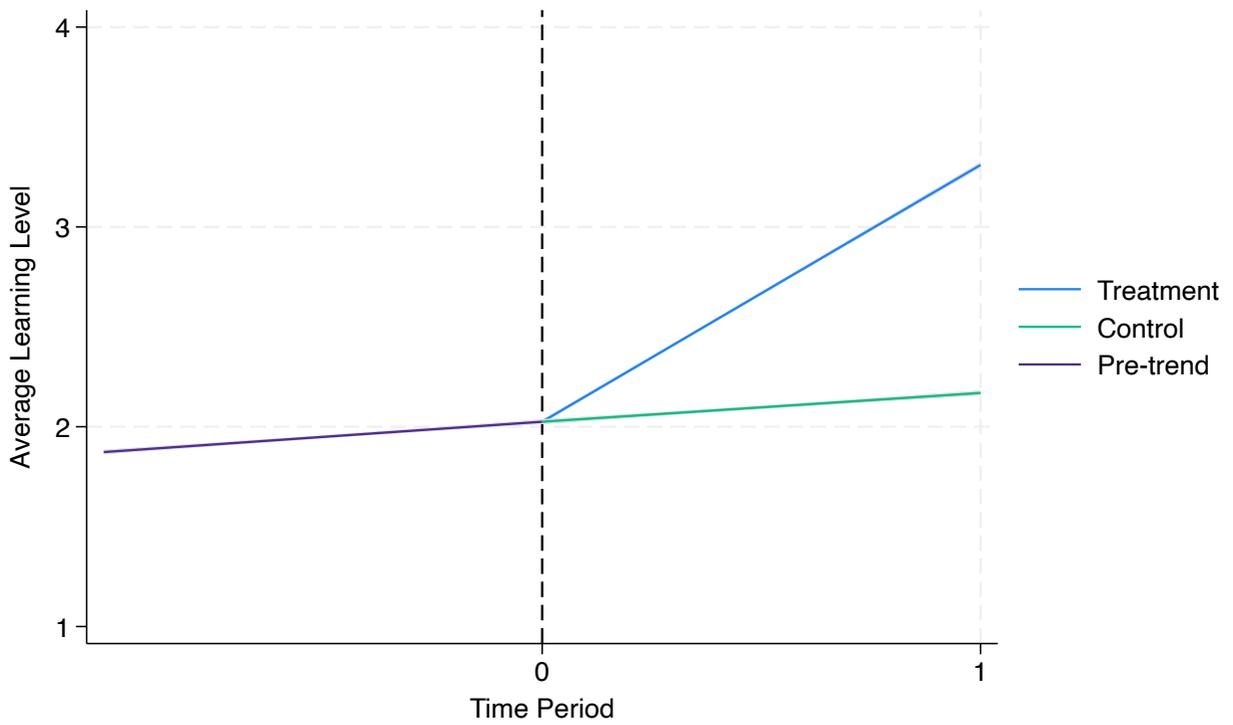
(e) India

Figure A7: COVID-19 Infections and School Disruption



Notes: This figure shows the evolution of the COVID-19 cases (in terms of new cases per million people) and resulting school closures or disruptions from April 2020 to April 2022. Each panel represents a country in our 5-country study, with the blue line indicating the approximate baseline collection date and the red line indicating the approximate endline collection date. Data sources include the Our World in Data COVID-19 dataset and UNESCO Global monitoring dataset of school closures

Figure A8: Difference-in-Difference Effects at Scale in India



Notes: This figure visualizes the difference-in-difference estimation. We show average learning levels for grade 4 students exposed to treatment as well as average learning trajectories in the status quo estimated by baseline differences in learning levels between grades 4 and 5. Grade 4 is coded as the 0 year and grade 5 time period 1. Grade 3 baseline levels are coded as the -1 year, since baseline differences between grade 3 and 4 enable estimation of trends in prior years. These learning levels are based on data for over 25,000 teachers and using data from around 100,000 children. Average learning levels are coded as follows: 1 = addition; 2 = subtraction; 3 = multiplication; 4 = division.

C Survey Tools

Building Resilient Education Systems: Baseline Questionnaire

1. Did you speak to the household?

- Yes
 No

2. Did the child's caregiver provide consent to participate?

- Yes
 No

3. If no consent was given, why?

4. What is the student's name and surname? _____

5. In what school is the student enrolled? _____

6. In what district/region is the school located? _____

7. What class/grade is the student currently enrolled in? _____

8. What is the student's age? *[Note: This question was asked in India, Uganda and the Philippines.]*

9. What is the student's gender? _____

MATH LEARNING MODULE: PROTOCOL

- Inform the caregiver that you would like the children to work on math problems.
- Ask the caregiver to place the call on speakerphone. They may repeat questions for the children to answer.
- Request that children answer math problems on their own on a scrap paper, including their answer. Children should work alone and not copy off anyone or work together.
- Explain that this is not an exam/test, so it's okay if the child(ren) do not get the answers correct.
- Children should take no longer than (30 seconds for Place Value) or (2 minutes for regular operation) to answer the question. If it seems someone is helping, gently ask them to refrain. If someone continues to help and/or the child takes longer than (30 seconds for place value) or (2 minutes for operations), mark that the child got this wrong.
- When finished, request that the child read out each of their answers and explain their answer to you. This is not an exam/test, so it's okay if your child(ren) does not get the answers correct. Are the child(ren) ready?
- Record all correct responses from each child and make a note of the highest operation that each child can perform.
- All students begin by answering the PLACE VALUE question and proceed with the LEARNING MODULE regardless of whether they answer with a correct response.

10. Prativa has 32 apples and organizes them by PLACE VALUE. How many TENS does she have?

- The child got the question correct
 The child got the question incorrect
 The child gives the correct answer but is not able to convincingly explain how they got their answer/I don't believe they answered it themselves.
 The parent was answering for the child/not letting the child answer, or the child used a calculator

11. The student solves: $34 + 47$

- The child got the question correct
 The child got the question incorrect
 The child gives the correct answer but is not able to convincingly explain how they got their answer/I don't believe they answered it themselves.
 The parent was answering for the child/not letting the child answer, or the child used a calculator

12. The student solves: 83 - 45

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or the child used a calculator

13. The student solves: 23*4 (23 multiplied by 4)

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or the child used a calculator

14. The student solves: 80/9 (80 divided by 9)

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or the child used a calculator

15. What languages is the learner able to speak? _____**16. What is the main caregiver's highest level of education? [Note: This question was asked at baseline in Nepal and Philippines and endline in other countries. Answer choices were adjusted to the contexts where necessary.]**

- None/ Informal Education
- Primary Education
- High School
- Certificate/ Vocational
- College/ Bachelor's Degree
- Master's Degree or higher
- Respondent does not know

17. What is the best phone number to reach the student? _____**18. Who is the owner of this phone (first name and surname)? _____****19. What is the relationship of the phone owner to the student? _____****20. Please list one alternate phone number where the child can be reached. _____**

Building Resilient Education Systems: Endline Questionnaire

1. Did you speak to the person that typically helps the child with math problems and schoolwork?

- Yes
- No

2. Did they provide consent to participate in the endline survey?

- Yes
- No

3. Who typically provides educational instruction or support to the child outside of school?

- Mother
- Father
- Grandparent
- Sibling
- Others not mentioned above in the household
- Teacher
- Adults from the community who do not live in the household

4. What is the main caregiver's highest level of education?

- None/ Informal Education
- Primary Education
- High School
- Certificate/ Vocational
- College/ Bachelor's Degree
- Master's Degree or higher
- Respondent does not know

5. How confident do you feel that [student name] made progress in learning over the past three months?

- The child would have significant difficulty performing any operation
- Not confident at all
- Slightly confident
- Moderately confident
- Very confident

6. What is the highest operation that you think the child can easily perform?

- The child would have significant difficulty performing any operation
- Addition
- Subtraction
- Multiplication
- Division
- Respondent does not know

7. Would you be interested in receiving phone-based education support in the future?

- Yes, through both phone calls and SMS
- Yes, only through SMS
- Yes, only through phone calls
- Yes, through either phone calls or SMS
- No, they are not interested

8. Over the past four weeks, how often did you spend time doing educational activities (in general) with [student name]? For example, reading, practicing math problems, composition, etc.

- Never
- Less than once per week
- 1-2 times per week
- 3-4 times per week
- 5 or more times per week

9. Over the past four weeks, how often did you spend time doing educational activities (in general) with children in your household OTHER than [student name]? For example, reading, practicing math problems, composition, etc.

- Never
- Less than once per week
- 1-2 times per week
- 3-4 times per week
- 5 or more times per week

10. Is the child currently in school?

- Yes - in person has been in the same school for the past 3 months
- Yes - in person switched to a new school in the past 3 months
- No - not in school currently because school is closed but will return when school is open
- No - not in school currently because school is closed and unlikely to return when schools reopen
- No - not in school currently. School is open but the student has not returned.
- F. Respondent refused to answer *[Note: These answer options were slightly adjusted to suit the context if necessary.]*

11. Were SMS messages that had math problems sent to you and [student name] over the past [context-specific time frame]?

- Yes
- No
- Respondent does not know

12. Over the past [context-specific time frame], how often did the child practice math problems that were sent to you in an SMS message?

- Always
- Frequently
- Sometimes
- Rarely
- Never
- The respondent does not know

13. How many phone calls for math instruction did the child receive over [context-specific time frame]?

Integer

14. Is the person that typically helps the child with their math problems / schoolwork currently working?

- Yes, full-time
- Yes, part-time
- No, they have retired
- No, they are unemployed

15. How true is the following statement about the child? - The child has many worries, or often seems worried.

- Not true
- Somewhat true
- Definitely true

16. Hypothetically, how much would you be willing to pay for your child to receive weekly 1-on-1 math phone calls to support their learning?

- Nothing
- A little
- A lot

17. How much does [student name] enjoy school?

- Very much
- Somewhat
- Not much
- They do not like school at all

18. Which maths problems does the child say they can easily perform?

- The child would have significant difficulty performing any operation
- Addition
- Subtraction
- Multiplication
- Division
- Respondent does not know

CORE LEARNING MODULE PROTOCOL

- Inform the caregiver that you would like the children to work on math problems.
- Ask the caregiver to place the call on speakerphone. They may repeat questions for the children to answer.
- Request that children answer math problems on their own on a scrap paper, including their answer. Children should work alone and not copy off anyone or work together.
- Explain that this is not an exam/test, so it's okay if the child(ren) do not get the answers correct.
- Children should take no longer than (30 seconds for Place Value) or (2 minutes for regular operation) to answer the question. If it seems someone is helping, gently ask them to refrain. If someone continues to help and/or the child takes longer than (30 seconds for place value) or (2 minutes for operations), mark that the child got this wrong.
- When finished, request that the child read out each of their answers and explain their answer to you. This is not an exam/test, so it's okay if your child(ren) does not get the answers correct. Are the child(ren) ready?
- Record all correct responses from each child and make a note of the highest operation that each child can perform.
- All students begin by answering the PLACE VALUE question and proceed with the LEARNING MODULE regardless of whether they answer with a correct response.

19. Prativa has 47 apples and organizes them by PLACE VALUE. How many TENS does she have?

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

20. The student solves: $52 + 39$

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

21. The student solves: $42 - 29$

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

22. The student solves: 28×3 (28 multiplied by 3)

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

23. The student solves: $65/8$ (65 divided by 8)

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

24. A man drives 24km. Then he drives 17km. How many km did he drive in total?

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

25. I will now ask you a word riddle. If the day before yesterday was Tuesday, what day is it?

- The student got the answer correct
- The student did not get the answer correct within 45 seconds.

26. Would you like to try to answer another logic question?

- Yes, I'd like to try with a question as difficult as this one
- Yes, but I'd like to try an easier question
- No

27. The student solves: $3/8 + 4/8$ (three eighths plus four eighths)

- The child got the question correct
- The child got the question incorrect
- The child gives the correct answer but is not able to convincingly explain how they got their answer/ I don't believe they answered it themselves.
- The parent was answering for the child/not letting the child answer, or child used a calculator

NOTE ON ROBUSTNESS CHECKS

Multiple learning robustness tests were conducted, both within and separate to the surveys. These included: survey

backchecks where a random subset of households were re-surveyed on learning outcomes (India); students being randomly assigned to answer different math problems (with different numerals) of the same difficulty level (Nepal); an in-person assessment of the same math problems (Kenya); and the above listed effort riddle question to measure real student effort, distinct from learning. The results of these robustness tests are shown in Table 6.

D Additional Training and Implementation Details

Trainings were primarily conducted virtually. First, a training of trainers was conducted. Program coordinators and senior project staff from partner organisations in each country (government, NGOs, and the World Bank) took part in a country-specific practical, interactive 1-day Training of Trainers workshop. This was delivered online by staff from Youth Impact. This was followed by a 1-2 day training of teachers and tutors led by each country’s respective implementing organisations. These training sessions were contextualized based on country context, with training delivered in the local language and with contextually relevant insights and best practices. Implementer training was delivered online in all countries except in Kenya and Uganda, where training was in-person.

During implementation, Youth Impact’s master trainer and program coordinator had periodic calls with each implementing organization to help troubleshoot and share tips and best practices learned from implementation in other countries. For example, Youth Impact reviewed data with implementing partners, shared tips about the best times to schedule sessions with households throughout the week, and shared suggestions from how other countries had paced their lessons and targeted instruction successfully.

The government-led interventions were led largely by the government. The government assigned a point person within the ministry to lead the intervention and coordinate within relevant government structures and to support teachers. Government leads had regular phone and online communication with their teachers. To the extent that there was NGO involvement, it was in conducting the training of trainers and sharing of draft tools and phone call guidelines to ensure common content and approaches were delivered across treatments.

On content, the weekly messages included a set of math problems that students were encouraged to solve between calls. Each week, there was a problem provided for each level: i.e. an addition problem, a subtraction problem, a multiplication problem, and a division problem. A typical SMS message looked as follows:

“Welcome to Week 2! ADDITION: $14+46=?$; $18+33=?$; SUBTRACTION: He picked 32 apples and gave his friend 11. How many is he left with?; MULTIPLICATION: $23 \times 3=?$ $14 \times 2=?$; DIVISION: You need to divide 10 apples evenly between 2 friends. How many will each get?”

This approach enabled instructors to be able to adapt to the child’s level and pick from a menu of levelled content.

E Education in Emergencies Database References

Various sources were used to build the database for school disruptions from 2005 to 2023. This data is not meant to cover the universe of disruptions but rather to provide an illustrative snapshot of the degree to which school disruptions happen to start to capture their frequency. Our search included news articles, as well as reports from organizations such as Save the Children, UNICEF’s Children’s Climate Risk Index, ReliefWeb by the United Nations Office for the Coordination of Humanitarian Affairs (OCHA), and the Global Assessment Report on Disaster Risk Reduction. Major climate disruptions were identified through the international disaster database (EM-DAT). Below we list specific news sources documenting school disruptions.

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