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When bootstraps aren't enough: Demand, supply, and learning in a very low-income context

Alex Eble and Maya Escueta

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# When bootstraps aren't enough: Demand, supply, and learning in a very low-income context 

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#### Abstract

Can families in low-income contexts "pull themselves up by their bootstraps?" In rural Gambia, caregivers with high aspirations for their children's future education and career, measured before the child starts school, invest substantially more in their children's education. Despite this, essentially none of these children are literate or numerate three years later. In contrast, a bundled supply-side intervention administered in these same areas generates large literacy and numeracy gains. Conditional on receipt of this intervention, children of high-aspirations caregivers are 25 percent more likely to achieve literacy and numeracy than those of low-aspirations caregivers. Our results show that even in very low-income contexts, greater caregiver aspirations for children can map onto substantially different child learning outcomes, but only in the presence of adequate complementary inputs.


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## 1 Introduction

Many families wish to provide better lives for their children than experienced by previous generations. One central lever families use to achieve this goal is education. Intergenerational educational mobility has been shown to correspond to economic mobility (Black et al., 2011; Chetty et al., 2014, 2017), particularly in low and middle income countries (Azam and Bhatt, 2015; Asher et al., 2018; Alesina et al., 2021). A series of theoretical and empirical studies has shown a strong linkage from one specific type of desire for the future, known as "aspirations," to both greater investment in education and higher educational outcomes (cf. Beaman et al. 2012; Bernard et al. 2014; Genicot and Ray 2017; Lybbert and Wydick 2018; La Ferrara 2019). ${ }^{1}$ It is not clear, however, whether this relationship - between desire for a better future for one's children, investment in their education, and greater educational outcomes - also holds in contexts where complementary inputs are absent, or of low quality.

In this paper, we study two core questions: first, if caregivers in low-income contexts want to raise their children's learning levels, how much learning can they bring about on their own? Second, how does this relationship change if we relax the constraint of very low-quality educational supply? To do so, we follow children and their caregivers in rural Gambia over a period of three years, beginning from the time immediately prior to the child starting primary school. We measure parent hopes for their child's future education and career, or "aspirations," at baseline. We use these as coarse measures of the family's desire to help the child towards a future that differs from what the family currently has. We estimate the mapping from baseline aspirations onto subsequent educational investment and learning. We contrast these relationships with the impact of a supplyside intervention which dramatically raises learning levels, and show how the aspirations-learning relationship in this context varies with a large increase in the quality of educational supply. We also show that the commonly-used test score standard deviation (SD) metric dramatically overstates the mapping from aspirations to learning when baseline levels of learning are extremely low. ${ }^{2}$

[^1]We use data from a census of families in 169 villages in the two central regions of The Gambia. The data track families who, at the time of a baseline survey, intended to enroll at least one of their children in the first grade, for the first time, in the fall of 2015. This baseline survey collected families' educational and career aspirations for these children. Families and children were then followed over three years, during which time data was collected on the child's school enrollment, school-related time use, and on the family's educational expenditure for the child. At endline, these children were administered one-on-one tests of basic reading and math skills. ${ }^{3}$

We focus on caregivers' educational and career aspirations for the child, as in La Ferrara (2019). We operationalize this with two indicator variables, capturing the aspiration to have the child go to university, and the aspiration for the child to work in an urban area - a proxy for higher income jobs, given that income in the city is so much higher than that in the countryside. At baseline, sixty percent of families wish to send their children to university, roughly 10 percentage points lower than found in recent studies of aspirations in Ethiopia and Somalia (Bernard et al., 2014; Kipchumba et al., 2021), and almost 40 percent lower than the proportion of caregivers in rural India who aspire to send their child to junior college (Attanasio et al., 2020). A similar proportion aspire that their child will work in an urban area.

We find that higher baseline aspirations map onto greater subsequent investment in the child's education. Late enrollment in school is a common problem in Sub-Saharan Africa (Glewwe and Jacoby, 1995; Bommier and Lambert, 2000), and enrollment of children in school at younger ages is a strong predictor of greater overall educational attainment (Nonoyama-Tarumi et al., 2010). Caregivers with greater educational and career aspirations for their children are between three and six percentage points more likely to enroll these children in school in the first two years of the study, as compared to children of caregivers with lower aspirations. In the final year of the study, when there are no enrollment differences, high aspirations caregivers spend significantly more money on cation and, in particular, reporting impact evaluations of educational interventions. For reference, see the numerous meta-analyses which collect and analyze these studies (Kremer and Holla, 2009; McEwan, 2014; Ganimian and Murnane, 2016; Glewwe and Muralidharan, 2016; Evans and Yuan, 2020).
${ }^{3}$ These were Early Grade Reading and Math Assessments, also known as "EGRA" and "EGMA" tests, respectively. See Platas et al. (2014) and Dubeck and Gove (2015) for details on their development, implementation, and limitations.
the child's education, and their children spend more time per day on school-related tasks.
Higher baseline aspirations also map onto higher endline test scores which appear, at first, to suggest substantially higher learning levels. Children whose caregivers report higher aspirations at baseline score 0.28-0.30 SDs better on a composite score of endline tests of basic reading and math ability, administered three years after the initial aspirations data were collected. The SD metric is a popular way to measure learning gains in studies of education, particularly in developing countries (Kremer and Holla, 2009; McEwan, 2014; Evans and Yuan, 2020). Put in the context of the hundreds of impact evaluations of educational interventions in developing country contexts covered by a series of recent meta-analyses ${ }^{4}$, an estimate of 0.30 SD lies between the 75 th and 90th percentile of effect sizes reported in these studies.

By comparing our effect size estimates to estimates of skill-based learning gains, we show that the SD metric dramatically overstates the relationship between baseline aspirations and subsequent learning. Specifically, using standard definitions of literacy and numeracy, we estimate a precise zero relationship between baseline aspirations and endline levels of either skill. We also find very small gains in other, lower-level skill attainment.

The difference between the conclusions implied by the SD measure and the absolute measures of skill acquisition results from the fact that, in our context, learning levels are compressed at the very left tail of possible values (Platas et al., 2014; Dubeck and Gove, 2015). This occurs despite our use of tests specifically designed to measure early skill acquisition, i.e., at this leftmost part of the skill distribution. From such a low baseline, even a very small absolute gain in test scores in this context translates into a large change in SD units; here, the 0.30 SD gain we measure translates to the child being somewhat more likely to master one additional rudimentary skill, such as differentiating which of two single-digit numbers is larger, or which of three words starts with a different sound. Such gains leaves essentially all children in this sample far from mastery of other basic skills necessary for literacy and numeracy - and expected of grade 2 students in The Gambia - such as reading simple words or calculating basic sums. ${ }^{5}$

[^2]We argue further that our estimates provide a likely upper bound on the status-quo relationship between family motivation to improve children's lives, as measured by aspirations, and learning outcomes in this context. This is because potential unobservable confounders - for example, unobserved wealth or family preferences - are most likely to be positively correlated with both the aspirations we study and educational outcomes (Bernard et al., 2014; Ross, 2019). Should such unobserved traits influence our estimates, the true relationship would be even smaller than what we measure.

Next, we show that higher aspirations do map onto a much greater likelihood of achieving literacy and numeracy when high-quality complementary inputs on the supply side are present. As reported in Eble et al. (2021), a highly-resourced, bundled supply-side educational intervention randomly assigned to be offered in some of these same villages yielded transformative learning gains. In this paper, we show that this large increase in the quality of educational supply dramatically changes the relationship between baseline aspirations and endline learning.

Conditional on their village being randomly assigned to receive the intervention, children whose caregivers have high baseline educational aspirations for the child are 25 percent more likely to reach literacy and numeracy at endline than children in these same intervention villages whose caregivers do not express these aspirations at baseline. This finding echoes the results of a large RCT in Tanzania documenting evidence of complementarities between multiple supply-side educational inputs in increasing learning (Mbiti et al., 2019). We then analyze how aspirations interact with the intervention in the acquisition of individual skills at varying levels of difficulty. We find that the intervention is a substitute for caregiver aspirations in children's acquisition of low-level skills, and a complement to caregiver aspirations in children's acquisition of higher-level skills.

Finally, we discuss two potential alternative explanations for our estimates of how aspirations and the quality of educational supply interact to generate learning. These are that aspirations merely capture unobserved child ability or household wealth, respectively, which also lead to level of the population being studied has also been found in US schools (Hill et al., 2008).
greater learning when the quality of educational supply increases. Unlike in our analysis of these relationships focusing on rural Gambian families not receiving the external intervention, we cannot use a bounding argument to capture how unobserved factors may contribute to this second set of results. This is because the intervention could either substitute for or reinforce the role of these factors. Instead, we investigate the likely magnitude of these contributions. For child ability, a wide range of factors - including the extremely low proportion of caregivers who have ever gone to school or are able to read, the fact that aspirations are measured prior to the child starting school, and the fact that even after children go to school, caregivers in such contexts often have highly inaccurate beliefs about child ability (Dizon-Ross, 2019) - make it exceedingly unlikely that caregivers aspirations are merely a proxy for child ability. For wealth, we show that our main results are robust to including interactions between aspirations and measures of wealth and caregiver education, indicating that the interaction between aspirations and the intervention is not driven by this alternative explanation.

Our paper makes two key contributions. First, we advance understanding of how the demandside and supply-side interact to generate learning in low-income contexts (cf. Jensen 2010; Glewwe and Muralidharan 2016; Muralidharan et al. 2019; Romero et al. 2020). We document that even in the face of severe income poverty, many families want to provide a better life for their children and act upon this desire by investing more in their children's education. We then show that these desires can map onto large gains in learning, but only when complementary inputs on the supply side are present. This adds to other recent evidence on the presence of complementarities in producing learning in low-income settings (Mbiti et al., 2019; Kerwin and Thornton, 2021). Second, we show that in settings characterized by very low baseline learning levels, the test score standard deviation - a metric for measuring learning used in hundreds of previous studies (McEwan, 2014; Ganimian and Murnane, 2016; Evans and Yuan, 2020) - can yield an erroneously positive conclusion about the relative importance of educational inputs. In light of this, we argue that skill-based learning measures should be preferred when assessing the efficacy of interventions and the relative importance of various educational inputs in contexts with very low baseline levels of learning (Platas et
al., 2014; Dubeck and Gove, 2015; Filmer et al., 2020). We also contribute to the growing body of work on the role of aspirations in education and development (cf. Dalton et al. 2016; Genicot and Ray 2017; Lybbert and Wydick 2018; Serneels and Dercon 2021).

Our paper proceeds as follows. In Section 2 we describe our setting and data, and our measures of learning and aspirations. In Section 3 we present our research design. Section 4 shows our estimates of the mapping from aspirations to learning in the status quo of rural Gambia, how this varies by which measure of learning is used, and bounds these estimates given potential unobserved factors. Section 5 shows how this relationship changes with a substantial increase in the quality of educational supply and explores alternative explanations for these results. Section 6 concludes.

## 2 Background, setting, and data

In this section, we describe the setting in which we work, the data we analyze, and describe how we measure learning and aspirations.

### 2.1 Setting

Our study takes place in small, rural settlements in the North Bank and Lower River regions of The Gambia. The Gambia is located in West Africa, with Senegal on its border to the north, east, and south, and the Atlantic Ocean to its west. ${ }^{6}$ It is a former British colony and served as a major hub for the trans-Atlantic slave trade. Its population is roughly two million people, and its geographic area covers roughly 11,300 square kilometers (CIA, 2019). It is also very income poor: per-capita GDP was estimated to be $\$ 716$ in 2018. The country's main sources of economic activity are agriculture, tourism, remittances, and foreign aid.

In addition to income poverty, the country's education levels are also very low. In 2013, the Demographic and Health Surveys estimated that only 26.7 percent of adults living in rural areas were literate, and roughly half of adults in these areas had never been to school (The Gambia Bureau of Statistics and ICF International, 2014). Other national assessments of children's reading

[^3]and math abilities have shown that even among children, learning levels are dramatically lower than in other countries in the region (Sprenger-Charolles, 2008).

Our study focuses on the population of children and families enrolled in the randomized controlled trial reported in Eble et al. (2021). The research took place in 169 villages in the two central regions of The Gambia, beginning with the universe of villages in these two regions who had between 10 and 300 households according to the 2013 national census. ${ }^{7}$ All villages which met the predetermined eligibility criterion of having at least 10 eligible children were enrolled. ${ }^{8}$ Children were eligible if, at the time of enumeration in early 2015 , i) they were between the ages of 6 and 8 , ii) they had not yet entered the first grade, and iii) their primary caregiver intended to enroll them in the first grade in the coming academic year. Because presence in this sample is conditional on the caregiver intending to enroll the child in school in the coming year, this means that aspirations measured among participants may differ from the population in these areas. When abstracting from our sample to the population of children in this age range in our study areas, we make the following assumption: the trajectory of literacy and numeracy skills among excluded children is unlikely to be dramatically better than of study participants (though it could be either similar, or worse). This stems from the fact that excluded children will enter school later than study children, and later school entry corresponds to worse academic outcomes in similar settings (Glewwe and Jacoby, 1995; Bommier and Lambert, 2000).

There were 4,518 children enumerated at baseline, 3,825 for whom we have endline test scores. For the sake of brevity, we focus on these 3,825 students in our analysis. ${ }^{9}$ In the next section, we describe the characteristics of these children and their families.

[^4]
### 2.2 Data

The data began with a census of children in study villages meeting the eligibility criteria described above. Data were collected from these children and their caregivers over the period from January 2015 to May 2018. Participants were enumerated in early 2015 and randomization occurred in late 2015. Villages randomly assigned to the intervention arm received a highly-resourced, bundled intervention providing after-school remedial education delivered by para teachers. This program began in early 2016 and continued until the beginning of May 2018. In these villages, para teachers from within the village or nearby were hired and trained to use scripted lessons to deliver afterschool, supplementary education for 12 hours per week over the course of the study, following the official Gambian curriculum as children progressed through school. These para teachers were regularly monitored with a focus on "coaching," that is, improving their instructional capacity and ensuring student learning. ${ }^{10}$

In Table 1, we present a few key demographic characteristics of the children in our sample overall and separately by the arm of the trial into which they were randomized. From here onward, we refer to children in villages randomized to not receive the intervention as the "status quo" group and those in villages randomized to receive the intervention as the "intervention" group. At baseline, more than three quarters of all children were being cared for by adults - whom we call caregivers - who had never been to school. ${ }^{11}$ This is lower than average levels in The Gambia (The Gambia Bureau of Statistics and ICF International, 2014), consistent with the fact that the areas in which the study took place are lower-income, more remote, and less well-served by the government than many others in the country. We observe a simple proxy for wealth: whether the floor, walls, and roof of the home are made of synthetic materials, also used in Eble et al. (2021), with roughly one quarter of households living in homes constructed entirely out of synthetic materials.

We also collect three types of data on educational investment. The first captures child en-

[^5]Table 1: Demographic characteristics

|  | (1) | (2) | (3) |
| :---: | :---: | :---: | :---: |
|  | All | Status quo | Intervention |
| Child is female | 0.50 | 0.51 | 0.48 |
| Child has five or more siblings | 0.41 | 0.43 | 0.39 |
| Caregiver can read simple sentence | 0.08 | 0.08 | 0.08 |
| Caregiver is not child's mother | 0.23 | 0.22 | 0.23 |
| Books found in house | 0.67 | 0.65 | 0.69 |
| Caregiver education |  |  |  |
| Never been to formal schooling | 0.76 | 0.77 | 0.76 |
| At least some primary education | 0.16 | 0.15 | 0.16 |
| At least some junior secondary education | 0.06 | 0.06 | 0.06 |
| At least some senior education, or more | 0.02 | 0.02 | 0.02 |
| Household wealth |  |  |  |
| House is made of all natural materials | 0.06 | 0.05 | 0.08 |
| House is made of partially synthetic materials | 0.68 | 0.68 | 0.68 |
| House is made of all synthetic materials | 0.26 | 0.28 | 0.24 |
| Observations | 3,825 | 2,045 | 1,780 |
| Joint F-statistic (p-value) |  | $\begin{gathered} 0.572 \\ (\mathrm{p}=0.684) \end{gathered}$ |  |

Table 1 note: this table presents select demographic characteristics for children in our sample, both overall (column 1) and then separately by the treatment status to which they were randomized (columns 2 and 3, respectively). The joint F-statistic is a test of the null that these variables together are not jointly predictive of the child's randomization status to the intervention (treatment) or status quo (control) group, clustering by trial-assigned clusters of contiguous villages. All variables in this table, except for the number of observations, are binary, with $0=$ No and $1=$ Yes.
rollment in school, collected at the end of each academic year. At the end of the third year, we also measure the child's school-related time use on an average weekday and, separately, caregiver expenditure on the child's education.

### 2.3 Measuring learning

The study measured learning via endline tests conducted in May and June of 2018. It used early grade reading and math assessments - EGRA and EGMA tests, respectively (Platas et al., 2014; Dubeck and Gove, 2015) - which were administered to all study children, one-on-one, as per the test guidelines. These tests are closely aligned with the Gambian national curriculum for grades 13, and versions of these tests have also been used as part of government teacher assessment efforts since 2007.

Each test is comprised of different "subtasks," i.e., skills. These skills capture different aspects of the process through which children acquire the skills to read with comprehension and calculate basic figures. In Table A. 1 we describe the subtasks/skills evaluated by each test. There are six subtasks within each test. As the number of the subtask rises, so does the level of difficulty: for example, reading subtask 1 focuses on a skill, letter sound identification, believed to be much less difficult than that evaluated in reading subtask 4, familiar word recognition. We provide the full test papers in Appendix A.

We generate three key measures of learning using these tests. First, as in Eble et al. (2021), we generate a composite score of overall performance on the two tests. This is calculated as the average of the correct proportion of answers on each test, itself generated the average correct proportion of answers on each subtask. We estimate both the difference in raw scores between groups, as well as the transformation of this difference into standard deviation units using Cohen's d, henceforth our "SD" measure.

Our second and third measures of learning draw on child performance on the individual subtasks within each test. Our second measure is comprised of standard, binary measures of literacy and numeracy derived from child performance on certain subtasks within each test. A child is assessed to be literate if they can read "with good fluency" ( 45 words per minute) and correctly answer at least $80 \%$ of reading comprehension questions. A child is assessed to be numerate if they can successfully identify missing numbers in a sequence (e.g., $2,4, \ldots, 8$ ) in at least $70 \%$ of the questions on the test, and correctly answer at least $80 \%$ of word problems. These definitions are the same used in Fazzio et al. (2021). Second, we show child performance on individual subtasks, as measured by the proportion of questions in that subtask answered correctly. A key advantage of the EGRA and EGMA tests is that they are designed to be able to measure even very early learning gains in the process of attaining literacy and numeracy. Measuring performance across these subtasks allows us to show detailed learning trajectories across a spectrum of skills, from the very earliest stages of learning to more advanced skills on the path to these two benchmark abilities.

No baseline tests of learning were conducted. This decision was made at the end of pre-trial
fieldwork in consultation with the Gambian Ministry of Basic and Secondary Education and other experts in the area. As we document, baseline learning levels are incredibly low in this area. Our fieldwork suggested that, because we are focusing on children who had not yet been to school at the time of baseline enumeration (and prior to randomization), baseline tests would have generated only a trivially small number of non-zero scores, and therefore the cost - both financial and in terms of the time of participants - greatly exceeded the likely benefit of these tests, and we proceed assuming that every child starts from a zero baseline learning level in terms of these skills. The very low levels of these skills that we measure at endline in the status quo group, after the vast majority of students have completed three years of primary schooling, further support this assumption.

### 2.4 Measuring aspirations

In the baseline survey, prior to randomization and also before the child would enter school for the first time, we asked the child's main caregiver for about their aspirations for the child's future. As in La Ferrara (2019), we focus on two types of aspiration. The first is the caregiver's aspirations for their child's highest level of educational attainment. The second is the caregiver's aspirations for their child's career in adulthood. To capture educational aspirations, we asked the child's main caregiver: "ideally, what is the highest level of education you would like [child name] to attain?" To capture career aspirations, we asked the caregiver: "when [child name] is 20 years old, what job do you hope [she/he] will be doing?"

These questions were designed to capture broader, latent variables about the caregiver's aspirations for the child's education and career, respectively. They were piloted prior to use, and are similar to those asked in other studies of aspirations in Ethiopia, India, and Somalia (Bernard et al., 2014; Attanasio et al., 2020; Kipchumba et al., 2021). Lybbert and Wydick's 2018 study of aspirations differentiates between "aspirational hope" and "wishful hope", arguing that the latter are characterized by a lack of a viable pathway to achieve them. In our study area in The Gambia, as in the Ethiopian, Indian, and Somalian contexts referenced above, few individuals are likely to go to college or university. Nonetheless, many caregivers hope that their children will do so, and
we follow this body of prior research in referring to responses to the two questions as capturing aspirations. We see them as measures of a broader latent trait encompassing the caregiver's motivation to provide a future for the child different from what the family currently experiences. As we show, they are also important predictors of the subsequent future-oriented investment behavior we study. ${ }^{12}$

In Table 2, we summarize caregivers' responses to these questions. In Panel A we show this for all children, and then for status quo and intervention children, respectively, calculating the difference between the two groups. In Panel B, we conduct a similar analysis for girls and for boys. In Column 1, we first show the proportion of all children whose caregiver expresses the relevant aspiration for the child at baseline. We see that, at baseline, roughly 60 percent of caregivers report aspirations for the child to go to university. This is slightly lower than levels recently recorded in rural Ethiopia (Bernard et al., 2014) and Somalia (Kipchumba et al., 2021). The question regarding caregiver aspirations for their child's career allowed the respondent to answer freely; responses that did not clearly fall into one of 14 listed categories were recorded as given and later coded. Given economic conditions in The Gambia and our initial analysis of this response data, we considered the following two broad categories of employment: working for the government and working in an urban area. Since more than 85 percent of respondents chose some type of work for the government, leaving little variation to study, we instead focus on aspirations to work in an urban area. ${ }^{13}$ This captures most jobs which require literacy and numeracy skills, and also takes in the fact that income in the city is much higher than in the countryside. Roughly 65 percent of caregivers express this aspiration for their child's career. The correlation between aspirations for the child to attend university and for the child to work in an urban area is 0.187 , indicating substantial independent variation between the two.

Comparing across groups, we see no difference in baseline aspirations between intervention

[^6]Table 2: Aspirations at baseline

| Panel A: Control vs. treatment | (1) | (2) | (3) | (4) |
| :---: | :---: | :---: | :---: | :---: |
|  | All | Status quo | Intervention | P -value |
| Highest level of education: university | 0.61 | 0.61 | 0.61 | 0.89 |
| Broad work categories |  |  |  |  |
| Job in urban area | 0.65 | 0.65 | 0.65 | 0.93 |
| Government job | 0.87 | 0.86 | 0.87 | 0.70 |
| Top jobs aspired to |  |  |  |  |
| Teacher | 0.26 | 0.26 | 0.27 | 0.89 |
| Work for government (no further detail) | 0.22 | 0.23 | 0.21 | 0.55 |
| Nurse | 0.15 | 0.15 | 0.15 | 0.94 |
| Observations | 3,825 | 2,045 | 1,780 | - |
|  | (1) | (2) | (3) | (4) |
| Panel B: Female vs. Male | All | Female | Male | P -value |
| Highest level of education: university | 0.61 | 0.60 | 0.63 | 0.06 |
| Broad work categories |  |  |  |  |
| Job in urban area | 0.65 | 0.67 | 0.64 | 0.06 |
| Government job | 0.87 | 0.89 | 0.84 | 0.00 |
| Top jobs aspired to |  |  |  |  |
| Teacher | 0.26 | 0.26 | 0.27 | 0.31 |
| Work for government (no further detail) | 0.22 | 0.20 | 0.24 | 0.00 |
| Nurse | 0.15 | 0.24 | 0.07 | 0.00 |
| Observations | 3,825 | 1,895 | 1,930 | - |

Table 2 note: this table summaries the aspirations data that caregivers report for children in the trial at baseline, both overall (column 1) and, in columns 2 and 3, separately by the group as described in the panel title and column heading. Column 4 presents the $p$-value of at-test for the null that the status quo and intervention (Panel A) and female and male (Panel B) aspirations levels are equal at baseline, using a regression of the variable on the panel group variable to generate this statistic. We clustered standard errors by trial-assigned clusters of contiguous villages. These results are robust to adding controls for caregiver education and household wealth.
and status quo children. We also find small - no more than three percentage point - differences between caregivers' aspirations for girl and boy children for our main two aspirations variables. These differences are not statistically significant at conventional levels.

We next describe how baseline aspirations correlate with other baseline characteristics that might predict educational investment and learning levels. In Table 3, we present conditional means of aspirations levels by a series of variables related to relative economic prosperity, household configuration, and caregiver education. Our correlate variables include: the child's gender; a indicator
variable that captures whether the child's caregiver has received any formal schooling; an indicator for whether the caregiver can read a simple sentence - in the spirit of the ASER literacy test (Pratham, 2012) - at the time of a baseline survey; an indicator for the household wealth variable described earlier (a binary variable capturing whether the home is made of synthetic materials); an indicator variable for whether the child has five or more siblings, meant to proxy for the number of children across which educational investments must be divided; and an indicator for whether there were any books found in the child's home during the baseline survey. These characteristics are all predetermined relative to our measurement of aspirations.

The child's caregiver is roughly ten percentage points more likely to express either of these aspirations for the child if the caregiver has previously been to school. We also see a large difference in baseline educational aspirations by whether the caregiver is literate or not (recall that only eight percent of caregivers are literate), and a smaller difference by the presence of books in the house. There is some evidence of a difference in baseline career aspirations by household wealth, but not in educational aspirations. We see no differences in baseline aspirations by the number of siblings or the identity of the caregiver.

As expected, there is some mapping from caregiver education and literacy to aspirations. Caregivers who have been to school and, separately, those who can read, are significantly more likely to hold high educational and career aspirations at baseline. Recall, however, that less than a quarter of caregivers have ever been to school, and less than 10 percent can read a simple sentence. The majority of caregivers with no formal schooling and who cannot read still hold these aspirations. Furthermore, the pairwise correlations between caregiver aspirations and caregiver education or literacy are small: the pairwise correlation between formal schooling and educational aspirations is 0.114 , and for career aspirations it is 0.104 . We see roughly similar correlations ( 0.130 and 0.108 , respectively) for caregiver literacy and aspirations.

We find no evidence of correlation between our rough proxy of wealth and either aspirations measure, nor is there significant pairwise correlation between wealth and caregiver education. This is in line with the notion that, in rural parts of The Gambia, higher levels of wealth are

Table 3: Correlates of aspirations

|  | $(1)$ <br> Aspires that <br> child will go to <br> university | $(2)$ <br> Aspires that child <br> will find work in <br> urban area |
| :--- | :---: | :---: |
| Child gender |  |  |
| Male | 0.63 | 0.63 |
| Female | 0.60 | 0.68 |
| P-value of difference | $(0.08)$ | $(0.04)$ |
| Caregiver education |  |  |
| Caregiver has been to school | 0.71 | 0.75 |
| Caregiver has never been to school | 0.58 | 0.63 |
| P-value of difference | $(0.00)$ | $(0.00)$ |
|  |  |  |
| Caregiver literacy | 0.83 | 0.83 |
| Can read simple sentence | 0.60 | 0.64 |
| Cannot read simple sentence | $(0.00)$ | $(0.00)$ |
| P-value of difference |  |  |
|  |  |  |
| Materials of home | 0.62 | 0.68 |
| Home made of synthetic materials | 0.61 | 0.65 |
| Home made of natural materials | $(0.78)$ | $(0.13)$ |
| P-value of difference |  |  |
| Number of siblings | 0.62 | 0.65 |
| Less than five | 0.63 | 0.67 |
| Five or more | $(0.61)$ | $(0.42)$ |
| P-value of difference |  |  |
|  |  | 0.66 |
| Books in house | 0.64 | 0.65 |
| Books found in house | $0.00)$ | $(0.81)$ |
| No books found in house |  |  |
| P-value of difference |  |  |

Table 3 notes: this table shows the conditional means of aspirations across the baseline characteristics labeled in the left-most column. We transform these baseline characteristics into binary variables, showing the conditional mean of the aspiration for both values of the characteristic, and the p -value for a t -test of the null that the aspiration in question is equal for those with each value of the baseline characteristic.
not necessarily predictive of greater education, particularly given the importance of farming and animal husbandry. In our later analysis of the relative importance of baseline aspirations on subsequent educational investment and learning gains, we add controls for these variables, isolating the relationship between the portion of aspirations orthogonal to these variables and our dependent variables.

## 3 Research design

In this section we describe the analyses we conduct to answer our core research questions. Our first research question asks how aspirations at baseline map onto later investment in education and endline learning levels for families in rural parts of The Gambia. To answer this research question, we estimate the following equation:

$$
\begin{equation*}
y_{i c}=\alpha_{0}+\alpha_{1} A_{t=0, i c}+\alpha_{2} X_{t=0, i c}+\eta_{r}+\varepsilon_{i c} \tag{1}
\end{equation*}
$$

This regresses $y_{i c}$, the outcome variable of child $i$ in cluster $c$, on $\alpha_{0}$, a constant; $A_{t=0, i c}$, the aspirations of the caregiver for child $i$ at baseline (i.e., when $t=0$ ); $X_{t=0, i c}$, a vector of predetermined variables for child $i$, measured at baseline (these include all the variables shown in Table 3); and $\eta_{r}$, a region fixed effect. We cluster our standard errors at the level of contiguous clusters of villages, $\varepsilon_{i c} .{ }^{14}$ Our main coefficient of interest is $\alpha_{1}$, which captures the mapping from baseline aspirations to subsequent outcomes, after controlling for the baseline characteristics listed in Table 3, such as gender, wealth, and caregiver education.

In these analyses, we use only data from the status quo group. This is because, as shown in Eble et al. (2021), the intervention group's subsequent educational investment and endline learning levels are affected by receipt of the intervention, confounding our ability to measure the status quo mapping from baseline aspirations to subsequent outcomes among children in this context.

Our second research question asks whether the relationship from baseline aspirations to subsequent learning levels, estimated in question 1 , changes with a dramatic improvement in the quality

[^7]of educational supply. In these analyses we study children in both the status quo and intervention groups. Here we also use ordinary least squares, regressing the outcome variable on a constant, the randomly assigned treatment status of the village in which the child was enumerated, baseline aspirations, and their interaction:
\[

$$
\begin{equation*}
y_{i c}=\beta_{0}+\beta_{1} T_{c}+\beta_{2} A_{t=0, i c}+\beta_{3} T_{c} * A_{t=0, i c}+\beta_{4} X_{t=0, i c}+\eta_{r}+\varepsilon_{i c} \tag{2}
\end{equation*}
$$

\]

Here $T_{c}$ is child $i$ 's treatment status, and $A_{t=0, i c}$ is again the relevant measure of aspirations for the child reported by their caregiver at baseline. Here too we cluster our standard errors by cluster of contiguous villages. Our main coefficient of interest is $\beta_{3}$; the sign and significance of this coefficient indicates whether the change in the quality of educational supply induced by the intervention changes the mapping from baseline aspirations to endline learning. Because the intervention is randomized and baseline aspirations are pre-determined, we do not include additional controls. Because our paper reports exploratory analysis of existing data, we did not register a pre-analysis plan (Olken, 2015; Lin and Green, 2016), though the analysis for the broader RCT was pre-specified and pre-registered (Boone et al., 2015).

## 4 Aspirations, educational investment, and learning

In this section, we show how baseline aspirations map onto subsequent educational investments and endline learning levels in the status quo group. The "investments" we consider are time and money. Time is measured by enrollment in school each year, as well as the proportion of time that the child spends on school-related tasks on a typical weekday in the final year of the study. The measure of monetary investment we use is the caregiver's educational expenditure on the child's education, also captured in the third and final year of the study. We study the mapping from baseline aspirations to endline learning levels first using the child's performance on the endline test. We then contrast this with estimates generated using standard measures of literacy and numeracy based on performance on these tests, as well as measures of the child's mastery of various specific reading and math skills. Finally, we bound our results by describing the likely sign of any potential
influence from unobserved factors on our estimates.

### 4.1 Aspirations and educational investment

We first characterize the mapping from baseline aspirations levels to subsequent educational investment. We present our estimates in Table 4; the outcome variables, named in the column headings, are educational expenditure in year three of the study, child time use in year three of the study, and enrollment in school in each of the three study years.

We find a positive and statistically significant mapping from both types of baseline aspirations to subsequent educational investments. For educational expenditure, we see that children whose caregivers hold higher educational or career aspirations for the child spend between 10 and 15 percent more per year on costs related to the child's education. ${ }^{15}$ We also find a statistically significant difference in the proportion of time on a typical weekday that the child spends on school-related tasks, with more time spent by children whose caregiver expressed high educational aspirations at baseline, though not for those expressing high career aspirations. The mapping from these aspirations to educational expenditure and enrollment are of the same order of magnitude as the estimated effect of the intervention-driven aspirations gain on educational investment measured in Bernard et al. (2014).

For a coarser measure of investment, enrollment in school in each of the three study years, we find that children whose caregivers have higher educational or career aspirations for the child at baseline are more likely to be enrolled in school in the first two years of the study. This pattern disappears in year three of the study, at which point almost all children are enrolled in school. Nonetheless, this early difference is important: delayed enrollment in school is a strong predictor of lower overall educational attainment (Nonoyama-Tarumi et al., 2010).

To better understand what other baseline characteristics predict early educational investment, we also present coefficients for other control variables. These show a few key patterns. First, the mappings from other factors to educational investment have a similar order of magnitude as do

[^8]Table 4: Estimating the mapping from baseline aspirations to subsequent educational investment in the status quo group

|  | (1) <br> Educational expenditure | (2) <br> School-related time use | (3) Enrolled in school, year 1 | (4) <br> Enrolled in school, year 2 | (5) Enrolled in school, year 3 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |  |  |
| Aspiration: child will go to college ( $\alpha_{1}$ ) | $\begin{gathered} 79.04 * * \\ (27.23) \end{gathered}$ | $\begin{gathered} 0.019^{* *} \\ (0.007) \end{gathered}$ | $\begin{gathered} 0.032 \\ (0.028) \end{gathered}$ | $\begin{aligned} & 0.052 * * \\ & (0.025) \end{aligned}$ | $\begin{gathered} 0.005 \\ (0.008) \end{gathered}$ |
| Wealth index high | $\begin{gathered} 109.04 * * \\ (40.22) \end{gathered}$ | $\begin{gathered} 0.003 \\ (0.008) \end{gathered}$ | $\begin{gathered} -0.009 \\ (0.025) \end{gathered}$ | $\begin{aligned} & -0.029 \\ & (0.018) \end{aligned}$ | $\begin{gathered} -0.002 \\ (0.012) \end{gathered}$ |
| Caregiver can read simple sentence | $\begin{gathered} 79.51 \\ (70.06) \end{gathered}$ | $\begin{aligned} & 0.028 * * \\ & (0.012) \end{aligned}$ | $\begin{aligned} & 0.066^{*} \\ & (0.036) \end{aligned}$ | $\begin{gathered} 0.061 * * \\ (0.026) \end{gathered}$ | $\begin{aligned} & 0.017 * \\ & (0.008) \end{aligned}$ |
| Books found in house | $\begin{gathered} 72.49 * * \\ (28.23) \end{gathered}$ | $\begin{aligned} & 0.011^{*} \\ & (0.007) \end{aligned}$ | $\begin{gathered} 0.047 * * \\ (0.020) \end{gathered}$ | $\begin{gathered} 0.045 * * * \\ (0.014) \end{gathered}$ | $\begin{gathered} 0.007 \\ (0.006) \end{gathered}$ |
| Child is female | $\begin{aligned} & -14.32 \\ & (21.49) \end{aligned}$ | $\begin{gathered} 0.005 \\ (0.008) \end{gathered}$ | $\begin{gathered} 0.020 \\ (0.015) \end{gathered}$ | $\begin{gathered} 0.005 \\ (0.023) \end{gathered}$ | $\begin{gathered} 0.009 \\ (0.008) \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{gathered} 611.36 \\ 1,862 \end{gathered}$ | $\begin{aligned} & 0.545 \\ & 1,908 \end{aligned}$ | $\begin{aligned} & 0.825 \\ & 1,937 \end{aligned}$ | $\begin{aligned} & 0.802 \\ & 1,909 \end{aligned}$ | $\begin{aligned} & 0.971 \\ & 1,908 \end{aligned}$ |
| Panel B: Career aspirations |  |  |  |  |  |
| Aspiration: child will work in urban area ( $\alpha_{1}$ ) | $\begin{gathered} 74.14^{* *} \\ (26.88) \end{gathered}$ | $\begin{gathered} 0.005 \\ (0.006) \end{gathered}$ | $\begin{gathered} 0.036 \\ (0.025) \end{gathered}$ | $\begin{gathered} 0.055^{* * *} \\ (0.020) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.005) \end{gathered}$ |
| Wealth index high | $\begin{gathered} 106.89 * * \\ (39.59) \end{gathered}$ | $\begin{gathered} 0.003 \\ (0.008) \end{gathered}$ | $\begin{aligned} & -0.010 \\ & (0.025) \end{aligned}$ | $\begin{aligned} & -0.030 \\ & (0.018) \end{aligned}$ | $\begin{aligned} & -0.002 \\ & (0.012) \end{aligned}$ |
| Caregiver can read simple sentence | $\begin{gathered} 81.99 \\ (68.16) \end{gathered}$ | $\begin{gathered} 0.030^{* *} \\ (0.012) \end{gathered}$ | $\begin{aligned} & 0.067 * \\ & (0.036) \end{aligned}$ | $\begin{gathered} 0.062 * * \\ (0.027) \end{gathered}$ | $\begin{gathered} 0.017 * * \\ (0.009) \end{gathered}$ |
| Books found in house | $\begin{gathered} 77.20 * * \\ (28.00) \end{gathered}$ | $\begin{aligned} & 0.012 * \\ & (0.007) \end{aligned}$ | $\begin{gathered} 0.049 * * \\ (0.019) \end{gathered}$ | $\begin{gathered} 0.048 * * * \\ (0.014) \end{gathered}$ | $\begin{gathered} 0.007 \\ (0.006) \end{gathered}$ |
| Child is female | $\begin{aligned} & -20.85 \\ & (21.58) \end{aligned}$ | $\begin{gathered} 0.004 \\ (0.008) \end{gathered}$ | $\begin{gathered} 0.018 \\ (0.015) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.024) \end{gathered}$ | $\begin{gathered} 0.009 \\ (0.008) \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{gathered} 617.54 \\ 1,862 \end{gathered}$ | $\begin{aligned} & 0.553 \\ & 1.908 \end{aligned}$ | $\begin{aligned} & 0.820 \\ & 1,937 \end{aligned}$ | $0.799$ | $\begin{aligned} & 0.973 \\ & 1,908 \end{aligned}$ |

Table 4 notes: this table reports the results of estimating Equation 1 using the outcome variable given in the column heading and with the type of baseline aspirations (educational or career) indicated in the panel heading. Dependent variables are labeled in the column headings and defined in the text. These analyses include only children in the status quo group. We report clustered standard errors in parentheses below each estimated coefficient. Observations vary by column because outcome variables were collected at different times and some children were missed in some periods. Results are robust to including only the smallest estimation sample. The full set of controls is as indicated in Section 3. ${ }^{*} p<0.10,{ }^{* *} p<0.05,{ }^{* * *} p<0.01$.
those for baseline aspirations. Second, the signs of these estimated relationships are as expected; for example, there is a statistically significant positive relationship between wealth and educational expenditure.

### 4.2 Aspirations and learning outcomes

We next characterize the relationship between baseline aspirations and endline learning levels, as measured by performance on the endline test, both in terms of the raw composite endline test score and standard measures of literacy and numeracy based on performance on given subtasks, as described in Section 2.2. We present these results in Table 5. In column 1 we show this for raw test scores. We find that, after controlling for baseline characteristics, children whose caregivers have high baseline educational aspirations for the child perform 3.3 points better than children of caregivers with low baseline education aspirations. For reference, children in the low aspirations group answering 15 percent of questions on the endline test correctly (i.e., their score is 15 points). The analog difference for children of caregivers with high and low baseline career aspirations, respectively, is 3.8 points. These differences are both highly statistically significant.

We plot the distribution of these scores, by aspiration group, in Figure 1. This shows that the high aspirations group's test score distribution first-order stochastically dominates the that of the low aspirations group for both types of aspiration. Kolmogorov-Smirnov tests of equality of distributions reject equality with $\mathrm{p}<0.001$ in both cases.

Using the common practice of transforming raw score differences into standard deviation units, the mapping from baseline caregiver aspirations to endline learning appears very large. For educational aspirations, the raw difference translates into a difference of 0.28 SD , and for career aspirations, it would be $0.30 \mathrm{SD} .{ }^{16}$ Comparing these values to the hundred of studies summarized in a series of recent meta-analyses of evaluations of educational interventions in such contexts (c.f. Kremer and Holla, 2009; McEwan, 2014; Glewwe and Muralidharan, 2016; Evans and Yuan, 2020), they lie between the 75th and 90th percentile of reported effect estimates.

Next, we show that when these differences are described in terms of skills gained, the mapping

[^9]Table 5: Estimating the mapping from baseline aspirations to endline learning in the status quo group

|  | (1) <br> Endline test score | (2) <br> Child is <br> literate | (3) Child is numerate |
| :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |
| Aspiration: child will go to college ( $\alpha_{1}$ ) | $\begin{gathered} 3.278 * * * \\ (0.910) \end{gathered}$ | $\begin{gathered} -0.001 \\ (0.002) \end{gathered}$ | $\begin{gathered} -0.001 \\ (0.004) \end{gathered}$ |
| Wealth index high | $\begin{aligned} & 1.863 * \\ & (1.019) \end{aligned}$ | $\begin{gathered} 0.001 \\ (0.002) \end{gathered}$ | $\begin{aligned} & -0.001 \\ & (0.004) \end{aligned}$ |
| Caregiver can read | $\begin{gathered} 5.985 * * * \\ (1.482) \end{gathered}$ | $\begin{aligned} & -0.001 \\ & (0.001) \end{aligned}$ | $\begin{aligned} & -0.002 \\ & (0.002) \end{aligned}$ |
| Books found in house | $\begin{gathered} 2.960 * * * \\ (0.725) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.001) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.003) \end{gathered}$ |
| Child is female | $\begin{aligned} & 1.877 * * \\ & (0.843) \end{aligned}$ | $\begin{aligned} & -0.002 \\ & (0.001) \end{aligned}$ | $\begin{gathered} 0.001 \\ (0.003) \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{gathered} 14.964 \\ 1,971 \end{gathered}$ | $\begin{aligned} & 0.001 \\ & 1,971 \end{aligned}$ | $\begin{aligned} & 0.006 \\ & 1,970 \end{aligned}$ |
| Panel B: Career aspirations |  |  |  |
| Aspiration: child will work in urban area $\left(\alpha_{1}\right)$ | $\begin{gathered} 3.792 * * * \\ (0.658) \end{gathered}$ | $\begin{gathered} 0.002 \\ (0.001) \end{gathered}$ | $\begin{gathered} 0.003 \\ (0.004) \end{gathered}$ |
| Wealth index high | $\begin{aligned} & 1.771 * \\ & (1.034) \end{aligned}$ | $\begin{gathered} 0.001 \\ (0.002) \end{gathered}$ | $\begin{aligned} & -0.001 \\ & (0.004) \end{aligned}$ |
| Caregiver can read | $\begin{gathered} 6.023 * * * \\ (1.468) \end{gathered}$ | $\begin{gathered} -0.001 \\ (0.001) \end{gathered}$ | $\begin{aligned} & -0.002 \\ & (0.002) \end{aligned}$ |
| Books found in house | $\begin{gathered} 3.179 * * * \\ (0.711) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.001) \end{gathered}$ | $\begin{gathered} 0.001 \\ (0.003) \end{gathered}$ |
| Child is female | $\begin{aligned} & 1.583 * \\ & (0.878) \end{aligned}$ | $\begin{aligned} & -0.002 \\ & (0.001) \end{aligned}$ | $\begin{gathered} 0.001 \\ (0.003) \end{gathered}$ |
| Comparison group mean | 14.604 | 0.000 | 0.004 |
| Number of observations | 1,971 | 1,971 | 1,970 |

Table 5 notes: this table reports the results of estimating Equation 1 using the outcome variable given in the column heading and with the type of baseline aspirations (educational or career) indicated in the panel heading. Dependent variables are labeled in the column headings and defined in the text. These analyses include only children in the status quo group. We report clustered standard errors in parentheses below each estimated coefficient. The scale of the endline test score is 0-100. Literacy and numeracy are indicator variables. The full set of controls is as indicated in Section 3. ${ }^{*} p<0.10,{ }^{* *} p<0.05,{ }^{* * *} p<0.01$.

Figure 1: Distribution of endline test scores for the status quo group, by baseline aspirations


Panel A: Educational aspirations


Panel B: Career aspirations

Figure 1 notes: this figure shows kernel density plots of endline test scores for children whose caregivers did (red dashed line) and did not (solid blue line) express the aspiration listed in the panel title at baseline. In these plots, we focus on children in the status quo group (that is, in villages assigned to not receive the intervention) and for whom we have a test score, comprising 1,971 observations. The vertical lines show the mean test score of the group whose distribution is plotted using the same width, color, and pattern of line. Kolmogorov-Smirnov tests reject the equality of the two distributions with $p \leq 0.001$ in each panel.
from baseline aspirations to endline learning appears much smaller or disappears entirely. We report results from estimating Equation 1 using literacy and numeracy as our outcome variables in columns 2 and 3, respectively, of Table 5. In both cases we estimate precise zeros, finding no evidence of a relationship between baseline aspirations and endline likelihood of achieving literacy or numeracy. The confidence intervals we generate can reject anything larger than a one percentage point difference in the likelihood of having attained either skill at endline as a result of having a caregiver with high educational or career aspirations at baseline.

The second way we measure the mapping of aspirations to subsequent skill acquisition uses test score gains in terms of the various skills each test evaluates. ${ }^{17}$ In Figure 2, we plot the mean percent of correctly answered questions for skill by baseline aspirations levels. This figure illustrates just how low endline skill levels are in the status quo group: regardless of baseline aspirations, these children can correctly answer fewer than 10 percent of questions for most higher-level math and reading skills, such as single-digit subtraction or the ability to read simple, familiar words such as "and" and "but." It also highlights the very small absolute difference in skills between low aspirations and high aspirations children. In Tables A. 2 and A.3, we show the regression equivalent of these comparisons, estimating Equation 1 using the subtask score as the dependent variable.

Even the large relative difference in test score standard deviations between children with low and high caregiver aspirations at baseline translates into a small absolute difference in endline reading and math skills seen through the lens of skill acquisition. One way to see this is that our estimates of the low aspirations - high aspirations difference in endline test scores suggest that children whose caregivers hold high educational or career aspirations for the child would be more likely to master one more basic math or reading skill than children whose caregivers do not. In math, for example, this would mean these children would be able to differentiate which of two numbers was larger. In reading, it would mean that these children would be able to differentiate which of three words started with a different sound. Another helpful reference is that literacy and numeracy begin to manifest when a child correctly answers roughly 60 to 65 percent of questions

[^10]Figure 2: Endline skill levels in the status quo group, by baseline aspirations


Panel B: Career aspirations
Figure 2 notes: this figure shows endline performance, by baseline aspirations level, on each of the individual subtasks of the EGRA and EGMA tests, respectively. Panel titles indicate the aspiration being studied. In these plots, we focus on children in the status quo group (that is, in villages assigned to not receive the intervention) and for whom we have a test score, comprising 1,971 observations. The subtasks listed on the x -axis are described in Table A. 1 and the full test papers are given in Appendix A.
on these tests. Re-examining the distributions in Figure 1 through this lens, very few children score near these levels, with vanishingly few at or above them. In short, essentially no children in the status quo group are even close to achieving either literacy or numeracy at endline.

As a result, we conclude that higher aspirations by themselves - even with the greater levels of investment that correspond to them - are unlikely to map onto a greater likelihood of literacy or numeracy for the vast majority of children in rural Gambia. Furthermore, because in such contexts the school curriculum normally advances to higher level skills as children progress in grade level, even if these children do not master the necessary lower-level skills (Pritchett, 2013; Muralidharan et al., 2019), these children are highly unlikely to ever achieve basic literacy or numeracy.

This finding builds on other work on how higher aspirations may not always lead to educational gains. Dalton et al. (2016) use a model to show that people can hold suboptimally high aspirations, such that if there exists an insurmountably large gap between the aspiration and the person's current state, the person may choose to invest very little. They refer to this state as "aspirations frustration" or "aspirations failure." Ross (2019) shows empirical evidence of this phenomenon in rural India. Leight et al. (2021) show that an intervention to raise aspirations in Ethiopia, similar to that studied in Bernard et al. (2014), has no measurable effect on either aspirations or investment. Here we have shown that even if higher aspirations do map onto greater investment in education, this does not correspond to meaningfully different learning trajectories in the status quo.

### 4.3 Measuring learning

Given the wide range of studies which use the SD metric to quantify learning gains (cf. McEwan, 2014; Ganimian and Murnane, 2016; Glewwe and Muralidharan, 2016), using effect sizes to compare the relative effectiveness of interventions or inputs across contexts is alluring and, in practice, common (Kremer and Holla, 2009; McEwan, 2014; Glewwe and Muralidharan, 2016; Evans and Yuan, 2020). Our findings in the previous section highlight another core result of our paper: in cases where learning levels are very low, using the test score SD metric can lead to incorrect conclusions about the importance of different learning inputs. This adds to prior work outlining the
psychometric issues with the comparability of different tests (Furr, 2021). ${ }^{18}$
Our study emphasizes two core difficulties with using the SD metric to compare learning gains across contexts characterized by low baseline levels of learning. The first is that at very low levels of learning, a small absolute gain can be a large relative gain. For example, as we show in Figure 2, the 0.30 SD gain is comprised of very small gains in skills. We estimate that the same increase in the percentage of questions answered correctly from a mean of 60 in the endline test score would reduce the magnitude of this difference, measured in SD terms, by a third. Second, as we show later in the paper, the gain in skills such as literacy or numeracy acquired from a given increment in test scores varies dramatically with the child's position in the skill distribution. While the 0.30 SD gain in learning we measure for the mapping from baseline aspirations to endline learning is larger than the vast majority of effect sizes measured in many hundreds of educational impact evaluations conducted in developing countries; at the starting point where almost all children in our status quo group find themselves, however, it would require many multiples of that gain to generate a meaningful change in literacy.

This difference in interpretation is important. Were we to have relied only on test score standard deviations, we would have concluded that aspirations are a powerful predictor of learning gains. Using literacy, numeracy, and individual skill gains, however, we see that higher aspirations correspond to no greater likelihood of achieving literacy or numeracy, and only meager skill gains, during a crucial three-year period in these children's lives. In the context of US schools, Hill et al. (2008) report a similar pattern of greater effect sizes at lower grade levels (and thus lower levels of baseline skill), further highlighting the difficulty of using the SD measure for cross-context comparisons. ${ }^{19}$

There are several available tools for analyzing skill acquisition instead of test score SDs. The appropriateness of each tool will naturally depend on the context and research question. For assess-

[^11]ment of low-level learning gains across contexts, the acquisition of early life skills such as familiar word reading or mastery of simple arithmetic tasks can be easily compared, particularly within the same language of assessment (Platas et al., 2014; Dubeck and Gove, 2015). For comparing across interventions or regions within a single national context, analysis of grade-specific skills, as per the expectations of the national curriculum, can be used (Muralidharan et al., 2019). For cross-context comparison of learning at higher levels of schooling, Filmer et al. (2020) propose a tool, learning-adjusted years of schooling (also known as LAYS), which allows for cross-context comparison of learning gains.

### 4.4 Bounding our estimates for the status quo group

We argue that our estimates are likely to provide an upper bound for the true relationship between caregiver educational and career aspirations, educational investment, and child learning for these areas of rural Gambia. Aspirations for education and employment are often positively correlated with other hard-to-measure or unobservable traits - such as caregiver wealth, education, or other tastes and preferences - that are also positively correlated with child educational investment and outcomes (Bernard et al., 2014; Ross, 2019). Any confounding from such sources would cause our estimates to be exaggerated, relative to the true relationship (Wooldridge, 2016). Therefore, unless there exists some other important, unobserved trait which is negatively correlated with these specific aspirations and positively correlated with educational investment and learning outcomes (or vice versa), our estimates are larger in magnitude than the true relationship. This further emphasizes our main point that higher educational and career aspirations alone are likely insufficient to reach higher learning levels in this, and perhaps similar contexts.

As described in Section 2.2, presence in our sample is conditional on the caregiver intending to enroll the child in school in the coming year. ${ }^{20}$ This means that aspirations may differ between the sample and the population of children in rural Gambia. Extrapolating to this latter group, we believe our estimates show a slightly different parameter. Specifically, our estimates of $\alpha_{1}$ in

[^12]Table 5 are an upper bound on what a child in this population might enjoy were their caregivers to possess high aspirations. Our argument rests upon the assumption that the children excluded from our study by the eligibility criterion of caregiver enrollment intent at baseline are likely to have either a similar or worse learning trajectory than those we included in the study. This premise is supported by multiple studies showing negative consequences in terms of learning and educational attainment stemming from late school enrollment in similar contexts (cf. Glewwe and Jacoby 1995; Bommier and Lambert 2000).

## 5 Consequences of increasing the quality of educational supply

In this section, we show how the mapping from aspirations to learning outcomes changes when a key constraint - that of low quality educational supply - is relaxed. We also provide evidence on how demand, as measured by aspirations, and supply interact to generate learning at different levels of skill.

### 5.1 Aspirations, learning, and educational supply

We first estimate how the mapping from baseline aspirations onto later learning outcomes changes when the quality of educational supply is dramatically improved by receipt of a bundled, supplyside intervention. In Figure 3, we plot the distribution of test scores among the four relevant groups - high and low aspirations children who did and did not receive the intervention, respectively. As in Figure 1, we show separate panels for educational and career aspirations. In both plots, and for both treatment assignments, we find that higher baseline aspirations map onto higher endline test scores.

We then estimate Equation 2 using data from the entire sample, i.e., both the status quo and intervention groups. We show these results in Table 6, using the three summary learning outcomes - standardized test scores, literacy, and numeracy - studied in Section 4. In Panel A, we show these results for educational aspirations; in Panel B, we show them for career aspirations.

Our core finding is that the mapping from baseline educational aspirations to endline literacy and numeracy is large and significant in the presence of high quality educational supply. As shown

Figure 3: Distribution of endline test scores, by baseline aspirations and receipt of intervention


Panel A: Educational aspirations


Panel B: Career aspirations
Figure 3 notes: this figure shows kernel density plots of endline test scores for children whose caregivers did and did not express the aspiration listed in the panel title at baseline, and within these groups. We plot these two distributions separately for children who were and were not resident at baseline in a village which was randomly assigned to receive the intervention (that is, both the status quo and intervention), as indicated in the figure legends. The vertical lines show the mean test score of the group whose distribution is plotted with the same width, color, and pattern of line. A total of All 3,813 observations in our estimation sample from Table 6 were used to generate these figures.

Table 6: How the mapping from baseline aspirations to endline learning changes in the presence of a large supply-side intervention

|  | (1) <br> Endline test score | (2) Child is literate |  |
| :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |
| Baseline aspirations x intervention ( $\beta_{3}$ ) | $\begin{gathered} 0.39 \\ (1.58) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.02) \end{gathered}$ | $\begin{aligned} & 0.04 * \\ & (0.02) \end{aligned}$ |
| Aspirations ( $\beta_{2}$ ) | $\begin{gathered} 3.65 * * * \\ (0.92) \end{gathered}$ | $\begin{gathered} -0.00 \\ (0.00) \end{gathered}$ | $\begin{gathered} -0.00 \\ (0.01) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 45.52 * * * \\ (1.74) \end{gathered}$ | $\begin{gathered} 0.23 * * * \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.17 * * * \\ (0.02) \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{aligned} & 14.96 \\ & 3814 \end{aligned}$ | $\begin{gathered} 0.00 \\ 3.814 \end{gathered}$ | $\begin{gathered} 0.01 \\ 3,813 \end{gathered}$ |
| Panel B: Career aspirations |  |  |  |
| Baseline aspirations x intervention ( $\beta_{3}$ ) | $\begin{aligned} & -2.44^{*} \\ & (1.32) \end{aligned}$ | $\begin{gathered} 0.03 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.01 \\ (0.02) \end{gathered}$ |
| Aspirations ( $\beta_{2}$ ) | $\begin{gathered} 3.86^{* * *} \\ (0.64) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.00) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.00) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 47.33 * * * \\ (1.68) \end{gathered}$ | $\begin{gathered} 0.25 * * * \\ (0.03) \end{gathered}$ | $\begin{gathered} 0.18 * * * \\ (0.02) \end{gathered}$ |
| Comparison group mean | 14.60 | 0.00 | 0.00 |
| Number of observations | 3,814 | 3,814 | 3,813 |

Table 6 notes: this table reports our estimates of the parameters in Equation 2 for the outcomes listed in the column headings. The panel titles indicate which baseline aspiration was used to generate the estimates shown. Coefficient estimates are reported according to the row title. We report clustered standard errors in parentheses below each estimated coefficient. Each panel x column "cell" corresponds to a separate regression. Comparison group means are calculated for those in the status quo group whose caregiver did not express the aspiration given in the column title at baseline. ${ }^{*} p<0.10,{ }^{* *} p<0.05,{ }^{* * *} p<0.01$.
in Figure 1, in the rural Gambian status quo, the mapping from higher baseline aspirations to greater endline test scores leaves the child very far from achieving literacy or numeracy. As we can see in both Figure 3 and Table 7, conditional on receiving the dramatic improvement in the quality of educational supply provided by the intervention, baseline educational aspirations lead to a four to six percentage point (or 24 to 26 percent) greater likelihood of attaining literacy and numeracy at endline. For career aspirations, however, we see a smaller and statistically insignificant relationship between baseline aspirations and literacy and numeracy, conditional on receiving the intervention.

Unlike in Section 4.4, it is much more difficult to bound our estimates in this section using the likely sign of effects from other contributing sources. This is because the intervention could either substitute for inputs from other sources that might be correlated with our measure of aspirations such as household wealth or unobserved child ability - or complement them. If these inputs were complements, our estimates would be an upper bound, as the isolated mapping from aspirations and the intervention would likely be smaller. If they were substitutes, our estimates would likely be lower bounds. Instead, we argue that the most important feature of this section is the difference, in terms of order of magnitude and significance, between our estimates of $\alpha_{1}$ and $\beta_{3}$.

The difference in magnitude and significance between these two parameters reveals how demand and supply contribute to the production of learning in this context. Our results in the previous section show that our very coarse measure of demand, by itself, is correlated with investment behavior, and the acquisition of some basic skills, but ultimately very little high-level learning. The far greater estimates we present in this section show that in the presence of adequate supply, our measure of demand is correlated with a substantially greater likelihood of children reaching both literacy and numeracy. This is due to the intervention moving out the frontier from which highaspirations families invest and push their children. In terms of raw points on the test, the intervention generates a gain that is more than an order of magnitude larger than the low aspirations-high aspirations differential and, in terms of literacy and numeracy, we estimate intervention-driven gains of of between 17 and 25 percentage points from a counterfactual case of essentially zero likelihood of demonstrating either skill. From this starting point, family aspirations do map onto
very large gains in the likelihood of possessing these crucial skills at the end of three years of school.

The key message we take from these findings is that even in a very low-income context, demand can matter for the production of learning if there is also adequate supply. In the next section, we pursue this analysis further by analyzing child performance on different subtasks, capturing mastery of the skills that lead up to and, later, comprise, literacy and numeracy.

### 5.2 Demand, supply, and skill acquisition

We next show how demand and supply interact to generate different levels of skill in this context. In Tables 7 and 8, we present estimates from using Equation 2 to regress the acquisition of individual skills - as measured by child performance on the different subtasks in reading and math on each test - on aspirations, the intervention, and their interaction. Here again we focus on $\beta_{3}$, which captures the interaction between baseline aspirations and the large change in the quality of educational supply caused by the randomly assigned intervention. We also show a new parameter, the "interaction mean," which is the predicted endline test score for high-aspirations children, conditional on their living at baseline in a village assigned to receive the intervention. We calculate this by adding $\beta_{2}$ and $\beta_{3}$ and present it along with a p -value of a test of the null that it is equal to zero. The magnitude and statistical significance of this parameter tells us whether, in intervention villages, the high-aspirations group demonstrates higher level of the skill in question than do children of the low aspirations group.

Our results show how the interaction of demand and supply in generating skill varies by skill difficulty. Specifically, for the acquisition of simpler skills the intervention appears to be a substitute for baseline aspirations in generating learning, while for more difficult skills, the two appear to be complements. In Panel A of both of these tables, we see a clear positive gradient between $\beta_{3}$ and the difficulty of the skill being tested. For low-level skills, such as letter and number recognition (reading and math subtask 1, respectively), we estimate a statistically significant negative interaction term. For higher-level skills, such as familiar word reading and two-digit addition and

Table 7: Demand, supply, and reading skill acquisition

|  | Subtask <br> 1 | Subtask 2 | Subtask 3 | Subtask <br> 4 | Subtask 5a | Subtask $5 b$ | Subtask 6 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |  |  |  |  |
| Baseline aspirations x intervention ( $\beta_{3}$ ) | $\begin{aligned} & -1.745 \\ & (1.936) \end{aligned}$ | $\begin{aligned} & -1.044 \\ & (2.057) \end{aligned}$ | $\begin{gathered} 2.322 \\ (1.885) \end{gathered}$ | $\begin{gathered} 3.418 \\ (2.223) \end{gathered}$ | $\begin{gathered} 3.191 \\ (2.122) \end{gathered}$ | $\begin{gathered} 4.264^{* *} \\ (2.075) \end{gathered}$ | $\begin{aligned} & -0.651 \\ & (2.273) \end{aligned}$ |
| Baseline aspirations ( $\beta_{2}$ ) | $\begin{gathered} 3.349 * * \\ (1.276) \end{gathered}$ | $\begin{gathered} 4.095^{* *} * \\ (1.272) \end{gathered}$ | $\begin{aligned} & 1.648^{*} \\ & (0.841) \end{aligned}$ | $\begin{gathered} 2.421 * * * \\ (0.841) \end{gathered}$ | $\begin{gathered} 2.400^{* * *} \\ (0.866) \end{gathered}$ | $\begin{aligned} & 1.226^{* *} \\ & (0.543) \end{aligned}$ | $\begin{gathered} 1.052 \\ (0.872) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 55.737 * * * \\ (2.143) \end{gathered}$ | $\begin{gathered} 24.628^{* * *} \\ (2.082) \end{gathered}$ | $\begin{gathered} 45.475 * * * \\ (1.874) \end{gathered}$ | $\begin{gathered} 57.575 * * * \\ (2.227) \end{gathered}$ | $\begin{gathered} 54.428 * * * \\ (2.251) \end{gathered}$ | $\begin{gathered} 42.227^{* * *} \\ (2.075) \end{gathered}$ | $\begin{gathered} 56.889 * * * \\ (2.436) \end{gathered}$ |
| Interaction mean $\left(\beta_{2}+\beta_{3}\right)$ P -value $\left[\beta_{2}+\beta_{3}=0\right.$ ] | $\begin{gathered} 1.604 \\ {[0.274]} \end{gathered}$ | $\begin{gathered} 3.051 \\ {[0.067]} \end{gathered}$ | $\begin{gathered} 3.970 \\ {[0.022]} \end{gathered}$ | $\begin{gathered} 5.839 \\ {[0.005]} \end{gathered}$ | $\begin{gathered} 5.591 \\ {[0.005]} \end{gathered}$ | $\begin{gathered} 5.490 \\ {[0.008]} \end{gathered}$ | $\begin{gathered} 0.401 \\ {[0.851]} \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{gathered} 37.820 \\ 3,683 \end{gathered}$ | $\begin{gathered} 37.261 \\ 3,683 \end{gathered}$ | $\begin{gathered} 25.238 \\ 3,683 \end{gathered}$ | $\begin{gathered} 30.705 \\ 3,683 \end{gathered}$ | $\begin{gathered} 29.915 \\ 3,683 \end{gathered}$ | $\begin{gathered} 21.682 \\ 3,683 \end{gathered}$ | $\begin{gathered} 31.135 \\ 3,683 \end{gathered}$ |
| Panel B: Career aspirations |  |  |  |  |  |  |  |
| Baseline aspirations $x$ intervention ( $\beta_{3}$ ) | $\begin{gathered} -4.123^{* *} \\ (1.566) \end{gathered}$ | $\begin{aligned} & -0.633 \\ & (1.940) \end{aligned}$ | $\begin{aligned} & -1.305 \\ & (1.641) \end{aligned}$ | $\begin{aligned} & -1.228 \\ & (1.736) \end{aligned}$ | $\begin{aligned} & -1.176 \\ & (1.621) \end{aligned}$ | $\begin{gathered} 0.503 \\ (1.804) \end{gathered}$ | $\begin{aligned} & -1.368 \\ & (2.296) \end{aligned}$ |
| Baseline aspirations ( $\beta_{2}$ ) | $\begin{gathered} 4.355^{* * *} \\ (0.878) \end{gathered}$ | $\begin{gathered} 3.504 * * * \\ (1.252) \end{gathered}$ | $\begin{gathered} 2.332 * * * \\ (0.562) \end{gathered}$ | $\begin{gathered} 2.482 * * * \\ (0.597) \end{gathered}$ | $\begin{gathered} 2.386 * * * \\ (0.582) \end{gathered}$ | $\begin{gathered} 1.664 * * * \\ (0.380) \end{gathered}$ | $\begin{gathered} 2.422 * * * \\ (0.685) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 57.332 * * * \\ (2.011) \end{gathered}$ | $\begin{gathered} 24.367 * * * \\ (2.097) \end{gathered}$ | $\begin{gathered} 47.727^{* * *} \\ (1.963) \end{gathered}$ | $\begin{gathered} 60.433 * * * \\ (2.071) \end{gathered}$ | $\begin{gathered} 57.114 * * * \\ (2.054) \end{gathered}$ | $\begin{gathered} 44.489 * * * \\ (2.076) \end{gathered}$ | $\begin{gathered} 57.382 * * * \\ (2.414) \end{gathered}$ |
| Interaction mean $\left(\beta_{2}+\beta_{3}\right)$ P-value $\left[\beta_{2}+\beta_{3}=0\right]$ | $\begin{gathered} 0.232 \\ {[0.857]} \end{gathered}$ | $\begin{gathered} 2.871 \\ {[0.043]} \end{gathered}$ | $\begin{gathered} 1.027 \\ {[0.504]} \end{gathered}$ | $\begin{gathered} 1.254 \\ {[0.441]} \end{gathered}$ | $\begin{gathered} 1.210 \\ {[0.424]} \end{gathered}$ | $\begin{gathered} 2.167 \\ {[0.220]} \end{gathered}$ | $\begin{gathered} 1.054 \\ {[0.625]} \end{gathered}$ |
| Comparison group mean | 37.656 | 37.404 | 25.632 | 31.591 | 30.752 | 22.279 | 30.365 |
| Number of observations | 3,683 | 3,683 | 3,683 | 3,683 | 3,683 | 3,683 | 3,683 |

Table 7 notes: this table shows results for estimating Equation 2 for children's scores on the individual reading subtasks; panel titles indicate which aspiration is being studied. The dependent variable in each column is the subtask listed in the column heading; subtasks are described in Table A.1. We report clustered standard errors in parentheses below each estimated coefficient. The tests are shown in their entirety in Appendix A, divided by subtasks and with subtask indicated at the top of each relevant block of questions. The possible values of each subtask score range from zero to 100 percent of questions answered correctly. ${ }^{*} p<0.10,{ }^{* *} p<0.05,{ }^{* * *} p<0.01$.

Table 8: Demand, supply, and math skill acquisition

|  | Subtask 1 | Subtask 2 | Subtask 3 | Subtask $4 \mathrm{a}$ | $\begin{gathered} \hline \text { Subtask } \\ 4 \mathrm{~b} \end{gathered}$ | Subtask 5a | Subtask 5b | $\begin{gathered} \hline \text { Subtask } \\ 6 \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |  |  |  |  |  |
| Baseline aspirations x intervention ( $\beta_{3}$ ) | $\begin{gathered} -5.304 * * \\ (2.273) \end{gathered}$ | $\begin{aligned} & -4.032 * \\ & (2.242) \end{aligned}$ | $\begin{aligned} & 3.385 * \\ & (1.729) \end{aligned}$ | $\begin{gathered} 0.660 \\ (1.955) \end{gathered}$ | $\begin{gathered} 2.873 \\ (2.194) \end{gathered}$ | $\begin{gathered} 1.619 \\ (1.880) \end{gathered}$ | $\begin{gathered} 7.364 * * * \\ (2.474) \end{gathered}$ | $\begin{gathered} -0.193 \\ (2.031) \end{gathered}$ |
| Baseline aspirations ( $\beta_{2}$ ) | $\begin{gathered} 7.462^{* * *} \\ (1.983) \end{gathered}$ | $\begin{gathered} 7.650 * * * \\ (1.827) \end{gathered}$ | $\begin{gathered} 2.276 * * \\ (0.894) \end{gathered}$ | $\begin{gathered} 3.431 * * * \\ (1.070) \end{gathered}$ | $\begin{gathered} 2.859 * * * \\ (0.788) \end{gathered}$ | $\begin{gathered} 3.070 * * * \\ (0.811) \end{gathered}$ | $\begin{gathered} 1.046 \\ (0.681) \end{gathered}$ | $\begin{gathered} 5.416 * * * \\ (1.131) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 49.649 * * * \\ (2.866) \end{gathered}$ | $\begin{gathered} 49.716 * * * \\ (2.786) \end{gathered}$ | $\begin{gathered} 41.106 * * * \\ (1.773) \end{gathered}$ | $\begin{gathered} 46.540 * * * \\ (2.128) \end{gathered}$ | $\begin{gathered} 56.763 * * * \\ (2.169) \end{gathered}$ | $\begin{gathered} 39.007 * * * \\ (1.611) \end{gathered}$ | $\begin{gathered} 47.004^{* * *} \\ (2.260) \end{gathered}$ | $\begin{gathered} 26.923 * * * \\ (1.972) \end{gathered}$ |
| Interaction mean $\left(\beta_{2}+\beta_{3}\right)$ P -value $\left[\beta_{2}+\beta_{3}=0\right]$ | $\begin{gathered} 2.158 \\ {[0.049]} \end{gathered}$ | $\begin{gathered} 3.618 \\ {[0.006]} \end{gathered}$ | $\begin{gathered} 5.661 \\ {[0.000]} \end{gathered}$ | $\begin{gathered} 4.091 \\ {[0.015]} \end{gathered}$ | $\begin{gathered} 5.732 \\ {[0.006]} \end{gathered}$ | $\begin{gathered} 4.689 \\ {[0.007]} \end{gathered}$ | $\begin{gathered} 8.410 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} 5.223 \\ {[0.002]} \end{gathered}$ |
| Comparison group mean Number of observations | $\begin{gathered} 64.822 \\ 3,682 \end{gathered}$ | $\begin{gathered} 57.450 \\ 3,682 \end{gathered}$ | $\begin{gathered} 35.478 \\ 3,682 \end{gathered}$ | $\begin{gathered} 36.491 \\ 3,682 \end{gathered}$ | $\begin{gathered} 32.851 \\ 3,682 \end{gathered}$ | $\begin{gathered} 25.710 \\ 3,682 \end{gathered}$ | $\begin{gathered} 24.955 \\ 3,682 \end{gathered}$ | $\begin{gathered} 34.343 \\ 3,682 \end{gathered}$ |
| Panel B: Career aspirations |  |  |  |  |  |  |  |  |
| Baseline aspirations x intervention ( $\beta_{3}$ ) | $\begin{gathered} -6.480^{* * *} \\ (2.026) \end{gathered}$ | $\begin{gathered} -7.478 * * * \\ (1.909) \end{gathered}$ | $\begin{gathered} -1.931 \\ (1.560) \end{gathered}$ | $\begin{aligned} & -3.558^{*} \\ & (2.002) \end{aligned}$ | $\begin{gathered} -2.731 \\ (2.184) \end{gathered}$ | $\begin{gathered} -2.445 \\ (1.700) \end{gathered}$ | $\begin{gathered} 2.114 \\ (2.418) \end{gathered}$ | $\begin{aligned} & -2.499 \\ & (2.000) \end{aligned}$ |
| Baseline aspirations ( $\beta_{2}$ ) | $\begin{gathered} 7.279 * * * \\ (1.713) \end{gathered}$ | $\begin{gathered} 8.805 * * * \\ (1.576) \end{gathered}$ | $\begin{gathered} 3.029 * * * \\ (0.749) \end{gathered}$ | $\begin{gathered} 5.531 * * * \\ (1.045) \end{gathered}$ | $\begin{gathered} 4.826 * * * \\ (0.847) \end{gathered}$ | $\begin{gathered} 3.500 * * * \\ (0.682) \end{gathered}$ | $\begin{gathered} 1.527 * * * \\ (0.533) \end{gathered}$ | $\begin{gathered} 4.058 * * * \\ (1.040) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 50.567 * * * \\ (2.694) \end{gathered}$ | $\begin{gathered} 52.063 * * * \\ (2.544) \end{gathered}$ | $\begin{gathered} 44.406 * * * \\ (1.877) \end{gathered}$ | $\begin{gathered} 49.237 * * * \\ (2.145) \end{gathered}$ | $\begin{gathered} 60.274 * * * \\ (2.405) \end{gathered}$ | $\begin{gathered} 41.558^{* * *} \\ (1.725) \end{gathered}$ | $\begin{gathered} 50.108^{* * *} \\ (2.517) \end{gathered}$ | $\begin{gathered} 28.376 * * * \\ (2.081) \end{gathered}$ |
| Interaction mean $\left(\beta_{2}+\beta_{3}\right)$ <br> P -value $\left[\beta_{2}+\beta_{3}=0\right]$ | $\begin{gathered} 0.799 \\ {[0.454]} \end{gathered}$ | $\begin{gathered} 1.327 \\ {[0.208]} \end{gathered}$ | $\begin{gathered} 1.098 \\ {[0.421]} \end{gathered}$ | $\begin{gathered} 1.973 \\ {[0.246]} \end{gathered}$ | $\begin{gathered} 2.095 \\ {[0.296]} \end{gathered}$ | $\begin{gathered} 1.055 \\ {[0.496]} \end{gathered}$ | $\begin{gathered} 3.641 \\ {[0.125]} \end{gathered}$ | $\begin{gathered} 1.559 \\ {[0.360]} \end{gathered}$ |
| Comparison group mean | 65.094 | 57.615 | 36.279 | 36.287 | 33.057 | 26.468 | 25.887 | 35.698 |
| Number of observations | 3,682 | 3,682 | 3,682 | 3,682 | 3,682 | 3,682 | 3,682 | 3,682 |

Table 8 notes: this table shows results for estimating Equation 2 for children's scores on the individual math subtasks; panel titles indicate which aspiration is being studied. The dependent variable in each column is the subtask listed in the column heading; subtasks are described in Table A.1. We report clustered standard errors in parentheses below each estimated coefficient. The tests are shown in their entirety in Appendix A, divided by subtasks and with subtask indicated at the top of each relevant block of questions. The possible values of each subtask score range from zero to 100 percent of questions answered correctly. ${ }^{*} p<0.10,{ }^{* *} p<0.05,{ }^{* * *} p<0.01$.
subtraction, the estimates become positive and increase in magnitude. Furthermore, the confidence intervals around the lower-level subtasks ( 1 and 2 in both reading and math) exclude the point estimates for the higher-level subtasks (4 and 5), and vice versa, even when the estimates themselves are not statistically significant. For the two most difficult subtasks - reading and math subtasks 5 b, capturing reading comprehension and the ability to perform two digit subtraction with borrowing, respectively - we estimate a large and statistically significant positive interaction term. ${ }^{21}$ This indicates that conditional on receipt of the intervention, children of high-aspirations caregivers are no more likely to perform better on this subtask than children of low-aspirations caregivers. Looking at the estimates from the most difficult subtask - subtask 5 b in both tests - we see that the interaction is statistically significant, large, and positive. This indicates that at higher levels of skill, aspirations and the intervention complement each other. The interaction mean, too, grows in both magnitude and statistical significance as the subtask level increases. This further supports our claim that aspirations and educational supply are substitutes at low levels of skills, and complements at higher levels.

This reveals an important feature of the way that aspirations map onto learning. In both the status quo and intervention group, high-aspirations families are trying to push out on the frontier of learning. Figure 2 and Tables A. 2 and A. 3 show that, in the status quo group, the largest differences in endline learning between the children of low- and high-aspirations caregivers occur at the lowest level skills: subtasks 1 and 2 in both reading and math. The intervention greatly advances that frontier, so that the largest learning differences we see in the status quo group are erased in the intervention group. In other words, the intervention greatly raises the platform from which highaspirations families are reaching, allowing differences in learning levels to appear between children of low- and high-aspirations caregivers in the intervention group in the most difficult subtasks, where before (Figure 2 and Tables A. 2 and A.3) there were none.

In Panel B of both tables, covering our estimates for career aspirations, we see evidence of the first pattern but not the second. Career aspirations and the intervention also appear to be substitutes

[^13]in the production of lower-level skills, but we do not find evidence that they are complements for higher level skills. This is consistent with the notion that career aspirations differ from educational aspirations in terms of how they are acted upon, as we saw in column 2 of Table 4.

### 5.3 Alternative explanations

We next address two potential alternative explanations for our key results, focusing on potential unobserved correlates of our measures of aspirations. The first is correlation between unobserved child ability and aspirations. The second is correlation between unobserved family wealth and aspirations.

There are several reasons why the correlation between unobserved child ability and aspirations is highly unlikely to be the main explanation for our results. First, in this context and, particularly, at the time that we measure aspirations, caregivers are highly unlikely to know whether the child is of high ability. At the time these data are collected, the child had not yet been to school. As DizonRoss (2019) documents, even after children enroll in school, caregivers in low-income contexts often have highly inaccurate beliefs about child ability. Furthermore, more than three quarters of the caregivers of the children in our sample have never been to school, and over 90 percent of them cannot read a short, simple sentence. As a result, it is highly unlikely that they are able to identify academic skill among their children at the time when baseline aspirations are measured, prior to the child's first ever enrollment in school. Second, the educational investment behavior of caregivers corresponds to aspirations in a way that is consistent with the investment response to an experimentally-generated increase in aspirations as measured in Bernard et al. (2014). Third, while career and education aspirations both predict subsequent investment behavior, they are only mildly correlated (correlation 0.18 ).

Finally, we examine how much aspirations vary across children within families, as a proxy for unobservable (to us) within-family differences in ability that may manifest as aspirations differences. There are 151 caregivers in our sample with more than one child who is enrolled in our study. In 92 percent of these cases, the caregiver expresses the same educational aspirations for
each child under their care. In 70 percent of these cases, the caregiver expresses the same career aspirations for each child under their care. This suggests that, in the vast majority of cases, our aspirations data capture family desires for their children's futures, rather than serving as a proxy for family beliefs about individual child ability.

Similarly, there are several reasons why it is highly unlikely that some broader, latent socioeconomic variable is what drives the interaction between baseline aspirations and the supply-side intervention. First, we see evidence of baseline educational aspirations leading to greater likelihood of literacy and numeracy in the presence of the intervention, but no such relationship for career aspirations. Second, we conduct a robustness test which estimates an alternative version of Table 6 after adding interactions between the intervention and household wealth, caregiver education, caregiver literacy, and the presence of books in the home. We present these results for baseline educational and career aspirations in Tables A. 4 and A.5, respectively. These show that the main patterns we observe in Table 6 are robust to the inclusion of these other predictors of a potential response to the presence of the intervention. In other words, for a reasonable set of observable controls, we show that there is still a residual in the learning outcomes we study to be explained by educational aspirations which is not explained by the interaction of the intervention and these other traits of the children and their families which also predict learning. As in Tables 7 and 8, our estimates for career aspirations show no evidence of positive interaction effects, underscoring the difference between educational and career aspirations.

## 6 Conclusion

In this paper we study a common desire that many families across the world have: the aspiration for their children to live better lives than those lived by previous generations. A common path for realizing this desire is through greater education. We characterize the relationship between caregiver aspirations for children, subsequent educational investment, and later learning in a very low-income context. We first document that many families in this context posses this desire and act on it, investing more in their children's education even in the face of the extremely high opportunity
cost of investment. We then show that this corresponds to a much greater likelihood of achieving literacy and numeracy by the end of the third grade, but only when complementary resources on the supply side are present. Our findings belie the notion that families in such contexts merely need to wish and try harder to "pull themselves up by their bootstraps" to realize these desires.

We also show that, when baseline learning levels are very low, the widely used measure of test score standard deviations can lead to incorrect conclusions about how educational inputs do or do not generate meaningful differences in learning. We find that, in this context, the test score standard deviation generated a spuriously optimistic assessment of the relationship between aspirations and learning outcomes. We argue that in such contexts with very low counterfactual learning levels, skill-based measures of learning gains may be more accurate in capturing the importance of various educational inputs and the impact of interventions.

Overall, our research highlights an important feature of the educational experience of children and their families in extremely resource-poor contexts such as the one we study. As is the case across the world, many caregivers in our sample wish to improve the life chances of their children and help them to reach a prosperous adulthood. We show that these caregivers expend dear financial resources to do so, both in terms of money and their children's time. These investments yield a statistically significant return in terms of the child's performance on literacy and numeracy tests, relative to that of their peers. Sadly, because counterfactual learning levels are so low in the rural Gambian status quo, these relative gains still leave children dramatically short of reaching literacy and numeracy, among the most crucial skills for reaching later economic productivity and participating in many spheres of society. With the presence of complementary inputs on the supply side, however, these same aspirations map onto far greater likelihood of the child being able to read with understanding and conduct basic arithmetic. For policy, this suggests that while the demand side can yield important learning gains in low-income contexts, substantial increases in the quality of educational supply will also be necessary to address the very low levels of learning in the many pockets of extreme poverty in the developing world.

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## Appendix

Table A.1: Test subtasks

| Reading |  | Math |  |
| :---: | :---: | :---: | :---: |
| Subtask | Example | Subtask | Example |
| 1 | Read a letter's sound (e.g., "eh" for e) | 1 | Read a number (e.g., 1, 5, 22) |
| 2 | Differentiate sounds (e.g., which word starts with a different sound: book, dog, or boy) | 2 | Choose the larger number (e.g., 7 or 5) |
| 3 | Read a made-up word (e.g., tob) | 3 | Complete a sequence (e.g., 246 _ ) |
|  |  | 4a | Simple addition (e.g., 3+2) |
| 4 | Read a familiar word (e.g., but) | 4b | Two- and three-digit addition (e.g., $38+26$ ) |
| 5a | Read a short passage | 5a | Simple subtraction (e.g., 5-3) |
| 5b | Answer questions on the passage's content | 5b | Two- and three-digit subtraction (e.g., 59-37) |
| 6 | Listen to a different short passage, answer questions on the passage's content | 6 | Solve a simple word problem read aloud |

Table A. 1 notes: this table describes the individual "subtasks" within the reading (EGRA) and math (EGMA) tests administered at endline. The full test papers are given in Appendix A; the relevant subtask number for each block of questions is indicated in the test papers.

Table A.2: Mapping of aspirations at baseline to endline performance on reading subtasks in the status quo group

|  | Subtask 1 | Subtask <br> 2 | Subtask 3 | Subtask 4 | Subtask 5a | Subtask 5b | $\begin{gathered} \hline \hline \text { Subtask } \\ 6 \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |  |  |  |  |
| High baseline educational aspirations ( $\alpha_{1}$ ) | $\begin{gathered} 3.132 * * \\ (1.311) \end{gathered}$ | $\begin{gathered} 3.664 * * * \\ (1.275) \end{gathered}$ | $\begin{aligned} & 1.516^{*} \\ & (0.854) \end{aligned}$ | $\begin{gathered} 2.380 * * * \\ (0.872) \end{gathered}$ | $\begin{gathered} 2.210^{* *} \\ (0.894) \end{gathered}$ | $\begin{aligned} & 1.267 * * \\ & (0.559) \end{aligned}$ | $\begin{gathered} 1.123 \\ (0.890) \end{gathered}$ |
| Comparison group mean | 11.592 | 25.741 | 3.744 | 3.729 | 4.371 | 2.028 | 4.309 |
| Number of observations | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 |
| Panel B: Career aspirations |  |  |  |  |  |  |  |
| High baseline career aspirations ( $\alpha_{1}$ ) | $\begin{gathered} 4.177 * * * \\ (0.850) \end{gathered}$ | $\begin{gathered} 3.172 * * \\ (1.253) \end{gathered}$ | $\begin{gathered} 2.292 * * * \\ (0.559) \end{gathered}$ | $\begin{gathered} 2.620 * * * \\ (0.598) \end{gathered}$ | $\begin{gathered} 2.383 * * * \\ (0.576) \end{gathered}$ | $\begin{gathered} 1.794 * * * \\ (0.347) \end{gathered}$ | $\begin{gathered} 2.343 * * * \\ (0.655) \end{gathered}$ |
| Comparison group mean | 10.884 | 25.949 | 3.295 | 3.499 | 4.183 | 1.671 | 3.494 |
| Number of observations | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 | 1,971 |

Table A. 2 notes: this table shows results for estimating Equation 1 for children's scores on the individual reading subtasks. We restrict our attention in this table to children in the status quo group. We report clustered standard errors in parentheses below each estimated coefficient. The dependent variable in each column is the subtask number listed in the column heading. Subtasks are described in Table A.1. The tests are shown in their entirety in Appendix A, divided by subtasks and with subtask indicated at the top of each relevant block of questions. The possible values of each subtask score range from zero to 100 percent of questions answered correctly. ${ }^{*} p<0.10$, ${ }^{* *} p<0.05,{ }^{* * *} p<0.01$.

Table A.3: Mapping of aspirations at baseline to endline performance on math subtasks in the status quo group

|  | $\begin{gathered} \hline \text { Subtask } \\ 1 \end{gathered}$ | $\begin{gathered} \hline \text { Subtask } \\ 2 \end{gathered}$ | Subtask <br> 3 | Subtask $4 \mathrm{a}$ | $\begin{gathered} \hline \text { Subtask } \\ 4 \mathrm{~b} \end{gathered}$ | $\begin{gathered} \hline \hline \text { Subtask } \\ 5 \mathrm{a} \end{gathered}$ | $\begin{gathered} \hline \hline \text { Subtask } \\ 5 \mathrm{~b} \end{gathered}$ | $\begin{gathered} \text { Subtask } \\ 6 \\ \hline \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Educational aspirations |  |  |  |  |  |  |  |  |
| High baseline educational aspirations ( $\alpha_{1}$ ) | $\begin{gathered} 6.804 * * * \\ (2.001) \end{gathered}$ | $\begin{gathered} 7.031 * * * \\ (1.842) \end{gathered}$ | $\begin{gathered} 2.104 * * \\ (0.922) \end{gathered}$ | $\begin{gathered} 3.072 * * * \\ (1.095) \end{gathered}$ | $\begin{gathered} 2.652 * * * \\ (0.776) \end{gathered}$ | $\begin{gathered} 2.770 * * * \\ (0.844) \end{gathered}$ | $\begin{gathered} 0.847 \\ (0.645) \end{gathered}$ | $\begin{gathered} 5.112 * * * \\ (1.143) \end{gathered}$ |
| Comparison group mean | 41.153 | 33.866 | 16.109 | 14.594 | 6.337 | 7.414 | 2.978 | 21.779 |
| Number of observations | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 |
| Panel B: Career aspirations |  |  |  |  |  |  |  |  |
| High baseline career aspirations ( $\alpha_{1}$ ) | $\begin{gathered} 6.670 * * * \\ (1.713) \end{gathered}$ | $\begin{gathered} 8.180 * * * \\ (1.564) \end{gathered}$ | $\begin{gathered} 2.941 * * * \\ (0.742) \end{gathered}$ | $\begin{gathered} 5.215 * * * \\ (1.026) \end{gathered}$ | $\begin{gathered} 4.660^{* * *} \\ (0.793) \end{gathered}$ | $\begin{gathered} 3.230 * * * \\ (0.652) \end{gathered}$ | $\begin{gathered} 1.428 * * * \\ (0.504) \end{gathered}$ | $\begin{gathered} 3.857 * * * \\ (1.037) \end{gathered}$ |
| Comparison group mean | 41.183 | 33.074 | 15.623 | 13.371 | 4.958 | 7.132 | 2.597 | 22.450 |
| Number of observations | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 | 1,970 |

Table A. 3 notes: this table shows results for estimating Equation 1 for children's scores on the individual math subtasks. We restrict our attention in this table to children in the status quo group. We report clustered standard errors in parentheses below each estimated coefficient. The dependent variable in each column is the subtask number listed in the column heading. Subtasks are described in Table A.1. The tests are shown in their entirety in Appendix A, divided by subtasks and with subtask indicated at the top of each relevant block of questions. The possible values of each subtask score range from zero to 100 percent of questions answered correctly. ${ }^{*} p<0.10,{ }^{* *} p<0.05$, ${ }^{* * *} p<0.01$.

Table A.4: How the mapping from baseline educational aspirations to endline learning changes in the presence of a large supply-side intervention, adding interactions with various other predictors of learning
$\left.\begin{array}{lccc}\hline \hline & & (1) & (2) \\ \text { Endline }\end{array} \quad \begin{array}{c}(3) \\ \text { Child is } \\ \text { literate }\end{array} \quad \begin{array}{c}\text { Child is } \\ \text { numerate }\end{array}\right]$

Table A. 4 notes: this table shows results for estimating Equation 2 after adding the interaction terms shown here. This is an analog to Panel A of Table 6, adding the interaction terms shown here to test whether, for a reasonable set of observable controls, there is still a residual in the learning outcomes we study to be explained by aspirations which is not explained by the interaction of the intervention and other traits of the children and their families which also predict learning. We report clustered standard errors in parentheses below each estimated coefficient. ${ }^{*} p<0.10$, ${ }^{* *} p<0.05,{ }^{* * *} p<0.01$.

Table A.5: How the mapping from baseline career aspirations to endline learning changes in the presence of a large supply-side intervention, adding interactions with various other predictors of learning

|  | (1) <br> Endline test score | (2) <br> Child is <br> literate | (3) <br> Child is numerate |
| :---: | :---: | :---: | :---: |
| Baseline career aspirations x intervention ( $\beta_{3}$ ) | $\begin{aligned} & -2.39^{*} \\ & (1.32) \end{aligned}$ | $\begin{gathered} 0.03 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.01 \\ (0.02) \end{gathered}$ |
| Baseline career aspirations x household wealth | $\begin{gathered} 1.65 \\ (1.48) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.02) \end{gathered}$ | $\begin{gathered} -0.02 \\ (0.02) \end{gathered}$ |
| Baseline career aspirations x caregiver has never been to school | $\begin{gathered} 1.77 \\ (1.95) \end{gathered}$ | $\begin{gathered} -0.03 \\ (0.02) \end{gathered}$ | $\begin{gathered} -0.04 \\ (0.03) \end{gathered}$ |
| Baseline career aspirations x caregiver can read simple sentence | $\begin{gathered} 3.93 \\ (3.13) \end{gathered}$ | $\begin{gathered} 0.06 \\ (0.04) \end{gathered}$ | $\begin{gathered} 0.01 \\ (0.05) \end{gathered}$ |
| Baseline aspirations x books in house | $\begin{gathered} 1.58 \\ (1.40) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.02) \end{gathered}$ |
| Baseline career aspirations ( $\beta_{2}$ ) | $\begin{gathered} 0.77 \\ (2.04) \end{gathered}$ | $\begin{gathered} 0.02 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.04 \\ (0.03) \end{gathered}$ |
| Household wealth | $\begin{gathered} -0.13 \\ (1.29) \end{gathered}$ | $\begin{gathered} 0.01 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.02) \end{gathered}$ |
| Caregiver has never been to school | $\begin{gathered} -1.61 \\ (1.67) \end{gathered}$ | $\begin{gathered} 0.03 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.02 \\ (0.03) \end{gathered}$ |
| Caregiver can read simple sentence | $\begin{gathered} 1.02 \\ (2.52) \end{gathered}$ | $\begin{gathered} -0.02 \\ (0.04) \end{gathered}$ | $\begin{gathered} -0.02 \\ (0.05) \end{gathered}$ |
| Books in house | $\begin{gathered} 0.79 \\ (1.19) \end{gathered}$ | $\begin{gathered} -0.01 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.00 \\ (0.02) \end{gathered}$ |
| Intervention ( $\beta_{1}$ ) | $\begin{gathered} 47.26 * * * \\ (1.67) \end{gathered}$ | $\begin{gathered} 0.25 * * * \\ (0.03) \end{gathered}$ | $\begin{gathered} 0.18^{* * *} \\ (0.02) \end{gathered}$ |
| Comparison group mean | 14.60 | 0.00 | 0.00 |
| Number of observations | 3,814 | 3,814 | 3,813 |

Table A. 5 notes: this table shows results for estimating Equation 2 after adding the interaction terms shown here. This is an analog to Panel B of Table 6, adding the interaction terms shown here to test whether, for a reasonable set of observable controls, there is still a residual in the learning outcomes we study to be explained by aspirations which is not explained by the interaction of the intervention and other traits of the children and their families which also predict learning. We report clustered standard errors in parentheses below each estimated coefficient. ${ }^{*} p<0.10$, ${ }^{* *} p<0.05,{ }^{* * *} p<0.01$.

Figure A.1: Regions of The Gambia and study area


Panel A: The Gambia's location in West Africa


Panel B: Study area with The Gambia
Figure A. 1 notes: this figure shows the location of our study area. In Panel A, we show a map of the continent of Africa with The Gambia shown within the red circle. In Panel B, we show a map of the Gambia, indicating the two regions where the study took place.

For Online Publication: Appendix A: Test papers

Test papers begin on next page

# Early Grade Math Assessment in The Gambia: Instructions for Enumerators and Children Response Form 

## General Instructions

It is important to establish a playful and relaxed relationship with the child through an initial talk on topics of interest to the child (follow the text in bold below). The child should perceive the assessment more as a game rather than an evaluation. It is important that you ONLY read aloud the text in bold, slowly and clearly, so that the child can understand the exercises.
$\therefore$ Good morning. My name is $\qquad$ . And you, what's your name? I like to $\qquad$ . And you, what do you like to do? Now that you have done some reading games with my colleague, let's do some Maths game. Throughout this exercise, you can answer in the language that you prefer. Is that ok? [wait until the child responds] Are you ready? [wait until the child responds] Let's start.

Assessment start time: hh: $\qquad$ mm


Thank you, let's move to the next task.

| Subtask 2. Number discrimination (PRACTICE) | (1) $x$ |
| :---: | :---: |
| - Look at these numbers. Say which number is bigger [the child can only be considered correct if he/she "says" the bigger number, pointing is not enough]. <br> 84 <br> $\checkmark \doteq$ [If the child answered 8 , say] Well done, 8 is bigger. Let's try another example. <br> $\times$ - [If the child did not answer 8, say] The bigger number is 8. [Point to 8] This is 8 . [Point to 4] This is 4.8 is bigger than 4 . Let's try another example. |  |
| Look at these numbers. Say which number is bigger. <br> $10 \quad 12$ <br> $\checkmark \doteq$ [If the child answered 12, say] Well done, 12 is bigger. Let's continue. $\mathbf{x}$ - [If the child did not answer 12, say] The bigger number is 12. [Point to 10] This is 10. [Point to 12] This is $\mathbf{1 2 .} \mathbf{1 2}$ is bigger than 10. Let's continue. |  |


| Subtask 2. Number discrimination (TEST) |  | (䦽 Page |  | (1) $x$ |
| :---: | :---: | :---: | :---: | :---: |
| - Look at these numbers. Say which number is bigger. [repeat for each item] |  |  |  | If the child makes 4 successive errors at any point, say "thank you", discontinue this subtask, mark below and move to the next subtask. <br> O If the child hesitates for 5 seconds, provide the answer and then point to the next item and say "Go on". Mark the item that you provided answer as incorrect. |
| es $(\checkmark) 1=$ Correct <br> $(\checkmark) 0=$ Incorrect or without answer <br> ( $\varnothing$ ) Mark with a circle the self-corrections if you already marked as incorrect. <br> $(\boldsymbol{\square})=$ Mark the final answer provided with a bracket $(\boldsymbol{\square})$. |  |  |  |  |
| 1 | $7 \quad 5 \quad 7$ | 1 | 0 |  |
| 2 | $11 \quad 24 \quad \underline{24}$ | 1 | 0 |  |
| 3 | $47 \quad 34 \quad \underline{47}$ | 1 | 0 |  |
| 4 | $58 \quad 49 \quad 58$ | 1 | 0 |  |
| 5 | $65 \quad 67 \quad \underline{67}$ | 1 | 0 |  |
| 6 | 947894 | 1 | 0 |  |
| 7 | 146153153 | 1 | 0 |  |
| 8 | 287534534 | 1 | 0 |  |
| 9 | 623632632 | 1 | 0 |  |
| 10 | 867965965 | 1 | 0 |  |
| es Exercise discontinued because the child made 4 successive mistakes. $\square$ |  |  |  |  |
| NA2: |  | NE2: |  |  |
| es Which languages did the child use in this task? (circle all answers that apply) |  |  |  |  |
| English Pulaar | Mandinka Olof | (please specify) |  |  |

Effective Intervention

| Subtask 3. Missing Number (PRACTICE) | Page 4 | © $\boldsymbol{x}$ |
| :--- | :--- | :--- |
| P1』 Here are some numbers. $\mathbf{1}, \mathbf{2}$ and 4, what number goes here [point to the |  |  | empty box]?


$\checkmark$ - [If the child answered 3, say] Well done, it's 3. Let's do another one.
$\times$ - [If the child did not answer 3, say] The number 3 goes here. Say the numbers with me [point to each number]. 1, 2, $\mathbf{3}$ and 4 . 3 goes here. Let's try another one.

P2 - Here are some numbers. 5, 10 and 15, what number goes here?

$\checkmark$ - [If the child answered 20, say] Well done, it's 20 . Let's continue
$\times$ - [lf the child did not answer 20, say] The number 20 goes here. Say the numbers with me [point to each number]. 5, 10, 15 and $\mathbf{2 0 . 2 0}$ goes here. Let's continue.


Thank you, let's move to the next task.

| Subtask 4a. Addition (level 1) |  |  |  |  | [1] | and |  | (1) 60 seconds |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| - Paper and pencil |  |  |  |  |  |  |  | Start the timer when you say "start". <br> When the timer reaches 0 , say "stop." <br> If the child makes 4 successive errors at any point, say "thank you", discontinue this subtask, mark below and move to the next subtask. <br> O If the child hesitates for 5 seconds, provide the answer and then point to the next item and say "Go on". Mark the item that you provided answer as incorrect. |
| - In these two pages there are some addition questions [glide hand from top to bottom on the two pages]. You should start here [point to the first problem]. I will use the timer and will tell you when to start and when to stop. Say the answer for each question. If you don't know an answer, move to the next problem. If you want, you can use this paper and pencil. Are you ready? [wait until the child responds and prepare to time] Start. |  |  |  |  |  |  |  |  |
| $(\checkmark) 1=$ Correct <br> $(\checkmark) 0=$ Incorrect or without answer <br> ( $\varnothing$ ) Mark with a circle the self-corrections if you already marked as incorrect. <br> $(\square)=$ Mark the final answer provided with a bracket $(\square)$. |  |  |  |  |  |  |  |  |
| 1 $3+2=(5)$ <br> 2 $1+3=(4)$ |  | 1 | 0 | $117+8=(15)$ |  | 10 |  |  |
|  |  | 1 | 0 | 12 | $4+7=(11)$ | 1 | 0 |  |
| 3 | $4+5$ = (9) | 1 | 0 | 13 | $7+5=(12)$ | 1 | 0 |  |
| 4 | $6+2$ = (8) | 1 | 0 | 14 | $8+6=(14)$ | 1 | 0 |  |
| 5 | $8+1$ = (9) | 1 | 0 | 15 | $9+8=(17)$ | 1 | 0 |  |
| 6 | $3+3$ ( 6 ) | 1 | 0 | 16 | $6+7=(13)$ | 1 | 0 |  |
| 7 | $7+3$ ( 10 ) | 1 | 0 | 17 | $8+8=(16)$ | 1 | 0 |  |
| 8 | $3+9=(12)$ | 1 | 0 | 18 | $8+5=(13)$ | 1 | 0 |  |
| 9 | $2+8=(10)$ | 1 | 0 | 19 | $10+2=(12)$ | 1 | 0 |  |
| 10 | $9+3=(12)$ | 1 | 0 | 20 | $8+10=(18)$ | 1 | 0 |  |
| The child used: |  |  |  |  |  |  |  |  |
|  | Fingers to count. |  |  |  |  |  |  |  |
|  | Paper and pencil. |  |  |  |  |  |  |  |
|  | Solved the question in his/her head. |  |  |  |  |  |  |  |
| Tick $\checkmark$ all answers that apply. |  |  |  |  |  |  |  |  |
| Time remaining on timer at completion (SECONDS): |  |  |  |  |  |  |  |  |
| Exercise discontinued because the child made 4 successive mistakes. $\square$ |  |  |  |  |  |  |  |  |
| NA4a: |  |  |  |  | NE4a: |  |  |  |
| Which languages did the child use in this task? (circle all answers that apply) |  |  |  |  |  |  |  |  |
| English Pular |  | din |  |  | Others (please specify) |  |  |  |

## Thank you, let's move to the next task.

| Subtask 4b. Addition (level 2) |  |  | Page 9 | (1) $x$ |
| :---: | :---: | :---: | :---: | :---: |
| $\square$ Paper and pencil |  |  |  | Skip this subtask if <br> the child scores zero <br> in level 1 Addition questions. <br> WIf the child makes 4 successive errors, say "thank you", discontinue this subtask, mark below and move to the next subtask. <br> O If the child uses an inefficient strategy (e.g. tick marks), ask the child "Do you know another way to solve the problem? If "no", move to the next item after 5 seconds. <br> © If the child does not provide answer in 30 , point to the next item and say "Go on". You may give additional 30 second if the child is still processing the question. |
| - Here are some addition questions [glide hand from top to bottom]. Tell me the answer for each question. If you do not know the answer, move to the next one. If you want, you may use this paper and pencil. Are you ready? [wait until the child responds] Start here [point to the first problem] |  |  |  |  |
| $(\checkmark) 1=$ Correct <br> $(\checkmark) 0=$ Incorrect or without answer |  |  |  |  |
| 1 | $13+6=(19)$ | 1 | 0 |  |
| 2 | $18+7=(25)$ | 1 | 0 |  |
| 3 | $14+25=(39)$ | 1 | 0 |  |
| 4 | $22+37=(59)$ | 1 | 0 |  |
| 5 | $38+26=(64)$ | 1 | 0 |  |
| 6 | 234+512= (746) | 1 | 0 |  |
| The child used: |  |  |  |  |
| Fingers to count. |  |  |  |  |
| Paper and pencil. |  |  |  |  |
| Solved the questio | in his/her head. |  |  |  |
| Tick $\checkmark$ all answers that apply. |  |  |  |  |
| Exercise discontinued because the child made 4 successive errors. $\quad \square$ |  |  |  |  |
| NA4b: |  | NE4b: |  |  |
| Which languages did the child use in this task? (circle all answers that apply) |  |  |  |  |
| English Pulaar | Mandinka Olof | hers (pleas | secify) |  |

Thank you, let's move to the next task.


## Thank you, let's move to the next task.



Thank you, let's move to the next task.


```
\checkmark [If the child answers 2, say] Well done, 2 children stayed in the
classroom. Let's continue.
x [If the child does not answer 2, Put 3 counters on top of the table and say]
Imagine that these counters are children. One of the children gets out of the
classroom. Show me the child getting out of the classroom. How many
children stayed in the classroom?
Well done, two children stayed in the classroom. Let's continue.
```



Which languages did the child use in this task? (circle all answers that apply)
English Pulaar Mandinka Olof Others (please specify) $\qquad$
Thank you, you did a good job. Now please return to your own classroom/you can go home.

| Which language(s) did you use to apply this test? (circle all answers that apply) <br> English Pulaar $\quad$ Mandinka Olof $\quad$ Others (please specify) |
| :--- |
| Assessment end time: |

[^14]
## Early Grade Reading Assessment in The Gambia: Instructions for Enumerators and Children Response Form

## General Instructions

It is important to establish a playful and relaxed relationship with the child that will be assessed through an initial talk on topics of interest to the child (see example below). Use this time to identify whether the child is comfortable with the national language you use. The child should perceive the assessment more as a game rather than an evaluation. It is important that you do not deviate from the guidelines and ONLY read aloud the text in bold, slowly and clearly, so that the child can understand the exercises.

- Good morning/afternoon. My name is $\qquad$ and I work at Effective Intervention. And you, what's your name? [wait until the child responds] How is your family? [wait until the child responds] When I am not at work, I like to $\qquad$ . And you? What do you most enjoy doing when you are not at school? [wait until the child responds]


## Verbal Consent

- Let me tell you why I am here today. I am working with a project of Effective Intervention. We came today to your school to do an exercise to help us better understand how children learn how to read and do mathematics, and you were chosen to help us.
- We would like to ask for your help. But you do not have to take part if you do not want to.
- We are going to play reading and mathematics games. I am going to ask you to read letters, words and a short story out loud. Then you will go to my friend/colleague sitting at the other side (point to the direction of the EGMA enumerator), and he/she will ask you to identify numbers, do some calculations and solve a few problems.
- Sometimes I will use this timer to time how long it takes you to complete some of the tasks. If you hear it beeps, please do not pay attention to it.
- This is NOT a test and it will not affect your grade at school.
- Once we begin, if you would rather not answer a question, that's all right.
- Can we start? [wait until the child responds]

If the oral consent is obtained, please tick: $\square$
If the oral consent is not obtained, please make a note on the student list.

[^15]$\square$ hh: $\qquad$ mm

| Subtask 1. Letter Sound Identification | Page 1 |
| :--- | :--- |
| $\bullet$ Here is a page with many English letter sounds. Please tell me the SOUNDS |  | of as many letters as you can- not the NAMES of the letters, but the SOUNDS.

For example, [Point to " $A$ "] this letter sound is /a/.
Let's practice. [Point to " T "] Tell me what letter sound this is.
$\checkmark \doteq$ [If the child read /t/, say] Very good, this letter sound is /t/.
$\times 8$ [If the child did not read /t/, say] This letter sound is /t/.
[Point to " $b$ "]. Now let's try another one. Tell me what letter sound this is.
$\checkmark \doteq$ [If the child read /b/, say] Very good, this letter sound is /b/.
$\mathbf{x} \dot{-}$ [If the child did not read $/ \mathrm{b} /$, say] This letter sound is $/ \mathrm{b} /$.
Have you understood? [wait until the child replies]
When I say "start", start here [point to the first letter], and read through the page [sweep finger across first line]. I will use this timer and will tell you when to stop. Point to each letter and read out loud the letter sound. Read as fast and the best you can. If there is a letter sound you can't read, move to the next one.

Put your finger on the first letter [make sure the child does so]. Are you ready? [wait until the child responds and prepare to time] You can start.
( / ) Mark any incorrect words with a slash ( / ).
(Ø) Mark with a circle the self-corrections if you already marked as incorrect.
( $\beth$ ) Mark the final letter read with a bracket $(\square)$.

## Examples: A T b

| 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| L | i | h | R | S | y | E | O | w | T |
| i | e | T | m | G | t | a | d | n | B |
| h | O | A | E | U | r | L | e | R | u |
| g | R | e | N | i | r | m | t | s | r |
| S | T | E | C | p | A | F | c | a | E |
| y | s | K | A | O | C | O | h | t | P |
| e | A | e | s | M | F | n | u | R | t |
| A | y | H | N | S | i | g | m | i | L |
| b | i | L | O | i | o | E | p | r | x |
| N | V | C | D | e | d | J | z | O | n |

## 60 seconds

Start the timer when the child reads the first letter. Stop the timer when the child reads the last letter.
© If the child hesitates for 3 seconds, read that letter and then point to the next letter and say "Continue". Mark the letter you read as incorrect.

* When the timer reaches 0 , say "stop."
m If the child does not provide a single correct response on the first line, say "Thank you!", discontinue this subtask, check the box at the bottom, and go on to the next subtask.

Time remaining on timer at completion (SECONDS):
$\square$
NA1:

[^16]| Subt | sk 2: Lett | Sound D | scrimina |  |  | x |  | (1) $x$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| In this exercise, you will listen to the English words that I read. I will read three words and one of them starts with a different sound. I will read twice. Tell me which one starts with a different sound. <br> For example: <br> "cat", "car", "hot"; "cat", "car", "hot" which one starts with a different sound? <br> [If the child answered "hot", say] Very good, "hot" starts with a different sound. <br> $\times$ - [If the child did not answer "hot", say] "cat", "car", "hot". "hot" starts with a different sound than "cat" and "car". <br> Now let's try again: <br> "light", "count", "learn"; "light", "count", "learn", which one starts with a different sound? <br> $\checkmark$ © [If the child answered "count", say] Very good, "count" starts with a different sound. <br> $\boldsymbol{x}$ - [If the child did not answer "count", say] "light", "count", "learn". "count" starts with a different sound than "light" and "learn". <br> Did you understand? [wait until the child responds] Are you ready? [wait until the child responds] Let's start. |  |  |  |  |  |  |  | 逝 If the child does not provide a correct answer in the first 5 items, say "Thank you!", discontinue this subtask, check the box at the bottom, and go on to the next subtask. <br> © If the child hesitates for 5 seconds, provide the answer. Mark the item that you provided answer as "no response". |
|  | $\begin{aligned} & 1=\text { Corre } \\ & 1 \\ & 0=\text { Incol } \\ & \text {. }=\text { No } \end{aligned}$ |  |  |  |  |  |  |  |
|  | $\begin{array}{r} \hline \ldots . . . \\ \hline \mathrm{wh} \\ d \end{array}$ | h one starts erent soun |  | Correct answer | Correct | Incorrect | $\begin{gathered} \text { No } \\ \text { response } \end{gathered}$ |  |
| 1. | book | dog | boy | [dog ] | 1 | 0 |  |  |
| 2. | like | eat | egg | [like] | 1 | 0 |  |  |
| 3. | do | get | go | [do] | 1 | 0 |  |  |
| 4. | say | pay | sad | [pay] | 1 | 0 |  |  |
| 5. | apple | candle | ant | [candle] | 1 | 0 | . |  |
| 6. | sun | red | run | [sun] | 1 | 0 |  |  |
| 7. | bag | ball | kick | [kick] | 1 | 0 |  |  |
| 8. | is | if | of | [of] | 1 | 0 |  |  |
| 9. | from | drum | drive | [from] | 1 | 0 |  |  |
| 10. | fly | good | food | [good] | 1 | 0 |  |  |
| es Exercise discontinued because the child had no correct answers in the first 5 items. $\square$ |  |  |  |  |  |  |  |  |
| NA2: |  |  |  |  | NE2: |  |  |  |

Thank you, let's move to the next task.


Thank you, let's move to the next task.

| Subtask 4. | iliar Wo | Reading |  | 四 |  | © 60 seconds |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| In this sheet, there are some English words. Read as many words as you can. Do not spell the words, but read them. <br> For example, [Point to the word "cat"] this word is "cat". <br> Let's practice. [Point to the word "mat"]. Read this word. <br> $\checkmark$ - [If the child answered "mat", say] Very good, the word is "mat". <br> $\times$ - [If the child did not answer "mat", say] This word is "mat". <br> Now let's try another one. [Point to the word "top"] <br> $\checkmark$ - [If the child answered "top", say] Very good, the word is "top". <br> $\times$ - [If the child did not answer "top", say] This word is "top". <br> When I say "start", start here [point to the first word], and read through the page [sweep finger across first line]. I will use this timer and will tell you when to stop. Point to each word and read out loud. Read as fast and the best you can. If there is one word you can't read, move to the next one. Put your finger on the first word [make sure the child does so]. Are you ready? [wait until the child responds and prepare to time] Start. |  |  |  |  |  | Start the timer when the child reads the first word. Stop the timer when the child reads the last word. <br> Э) If the child hesitates for 3 seconds, provide the word and then point to the next word and say "Continue". Mark the word that you provided as incorrect. <br> When the timer reaches 0 , say "stop." |
| ( / ) Mark any incorrect words with a slash (/). <br> (ø) Mark with a circle the self-corrections if you already marked as incorrect. <br> ( $\beth$ ) Mark the final word read with a bracket ( $\overline{\text { }}$ ). |  |  |  |  |  | em the child does not provide a single correct response on the first line (5 words), say "Thank you!", discontinue this subtask, check the box at the bottom, and go on to the next subtask |
| Example: <br> 1 <br> but | $\begin{array}{ccr} \text { cat } & \text { mat } & \text { top } \\ 2 & 3 \\ \hline \end{array}$ |  | 4 | 5 | (5) <br> (10) |  |
|  | time | in | the | also |  |  |
| make | no | its | said | where |  |  |
| came | very | do | after | long | (15) |  |
| water | run | all | for | paper | (20) |  |
| her | was | three | been | more | (25) |  |
| that | must | can | ear | it | (30) |  |
| jump | words | back | called | work | (35) |  |
| could | an | him | on | see | (40) |  |
| that | get | not | zip | what | (45) |  |
| you | if | their | teacher | when | (50) |  |
| Time remaining on timer at completion (SECONDS): |  |  |  |  |  |  |
| Exercise discontinued because the child had no correct answers in the first line. $\square$ |  |  |  |  |  |  |
| NA4: |  |  | NE4: |  |  |  |

Thank you, let's move to the next task.


Thank you, let's move to the next task.

| Subtask 6. Listening comprehension |  | (1) $x$ |  |  |  |  | (1) $x$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| - I am going to read you a short story aloud ONCE and then ask you some questions. Please listen carefully and answer the questions as best as you can. You can answer the questions in whichever language you prefer. Ready? [wait until the child responds] |  |  |  |  |  |  | Remove the passage from the child's view. <br> Do not allow the child to look at the passage or the questions. <br> If a child says" " don't know", mark as incorrect. |
| Demba was very sad when he lost one of his goats. He could not go to look for the goat, because he had to watch the other goats. Demba's grandfather helped and found the goat. Demba was very happy. |  |  |  |  |  |  |  |
| - Now I am going to ask you some questions related to the story: |  |  |  | Correct | Incorrect | No response |  |
| Why was Demba sad? <br> [He lost his goat; he could not go to look for it; he cannot see his goat] |  |  |  | 1 | 0 |  |  |
| Who helped to look for the goat? <br> [Demba's grandfather, his grandfather, grandfather] |  |  |  | 1 | 0 |  |  |
| Why was Demba happy? <br> [Grandfather returned with his goat; his goat is back; Grandfather found the goat, he sees/saw the goat etc] |  |  |  | 1 | 0 |  |  |
| Which languages did the child use in this task? (circle all answers that apply) |  |  |  |  |  |  |  |
| English | Pulaar Mandinka | Wolof | Others (please |  |  |  |  |

Thank you for doing this exercise with me. [Follow the instruction on the enumeration manual]

| Which language(s) did you use to apply this test? (circle all answers that apply) |  |  |  |  |
| ---: | ---: | ---: | ---: | ---: | :--- |
| English | Pulaar | Mandinka | Wolof | Others (please specify) |

Assessment end time: $\quad \ldots \quad \mathrm{hh}: \ldots \quad \mathrm{mm}$

[^17]
[^0]:    *Eble: Teachers College, Columbia University, eble @tc.columbia.edu. Escueta: Sanford School of Public Policy, Duke University, maya.escueta@duke.edu. We are grateful to the collaborators on the SCORE Trial - Chris Frost, Alpha Camara, Baboucarr Bouy, Momodou Bah, Maitri Sivaraman, Pei-Tseng Jenny Hsieh, Chitra Jayanty, Tony Brady, Piotr Gawron, Stijn Vansteelandt, Peter Boone, and Diana Elbourne - as well as our research team in the Gambia, particularly Lamin Janneh; see Eble et al. (2021). This paper reports analysis of data collected in that study, which was approved by the Institutional Review Board of the London School of Hygiene and Tropical Medicine, protocol number 8767. We received helpful input from Kimi Chan, Beezer Cheble, Sarah Cohodes, Esther Gehrke, Naureen Karachiwalla, Josh Merfeld, Kate Orkin, Phillip Ross, and seminar audiences at the Columbia Committee on the Economics of Education, the Columbia Population Research Center, the Gui2de/World Bank joint seminar, the New York Nanodev conference, and the RISE 2021 Annual Conference. This paper updates and replaces the previously circulated paper "Aspirations, Education, and Extreme Poverty." Eble acknowledges support from the National Academy of Education (NAEd) and the NAEd/Spencer Postdoctoral Fellowship Program. Keywords: aspirations; education; poverty; complementarity.

[^1]:    ${ }^{1}$ See also the great summary of empirical work on this topic in Fruttero et al. (2021).
    ${ }^{2}$ The SD metric has been used for measuring learning gains in several hundred studies in the economics of edu-

[^2]:    ${ }^{4}$ See those cited in the previous sentence and in Footnote 2.
    ${ }^{5}$ This inverse relationship between the learning contained in a given effect size estimate and the baseline learning

[^3]:    ${ }^{6}$ In Figure A.1, Panel A, we show a map of The Gambia's location on the African continent.

[^4]:    ${ }^{7}$ In Figure A.1, Panel B, we show a map of The Gambia indicating the regions in which these villages are located.
    ${ }^{8}$ There were 323 total villages to begin with. Of these, 113 had too few children to be eligible. The study excluded a further 41 of the remaining villages to create buffer zones between villages in order to ensure no potential for spillover between villages, i.e., caregivers of children in control villages instructing their children to walk into an intervention village and avail themselves of the intervention there.
    ${ }^{9}$ Baseline aspirations do not predict attrition at the endline test.

[^5]:    ${ }^{10}$ This intervention is described in greater depth in Eble et al. (2021).
    ${ }^{11}$ We focus on caregivers, as opposed to parents, because early fieldwork suggested that the most important person for the child's development is the primary person from whom the child receives their day-to-day care. This is often, but not always, the parent. In our data, roughly $75 \%$ of caregivers are mothers, $11 \%$ are grandmothers, and the rest are various other members of the household in which the child lives.

[^6]:    ${ }^{12}$ The aspirations we measure also differ importantly from expectations. In our pilot, we worked to choose language that differentiated between aspirations and expectations. In this work, however, we determined that we could not ask respondents about both expectations and aspirations without unacceptably large priming effects.
    ${ }^{13}$ This includes jobs such as doctor, nurse, judge, legal clerk, or politician, but not jobs like imam, farmer, or farm laborer.

[^7]:    ${ }^{14}$ This is the same level as the randomization in Eble et al. (2021).

[^8]:    ${ }^{15}$ Expenditures are shown in 2020 US dollars; this difference is between 75 and 90 Gambian Dalasis, converted at a rate of 51.71 dalasis per dollar.

[^9]:    ${ }^{16}$ Estimated using Cohen's $d$.

[^10]:    ${ }^{17}$ See Table A. 1 for a description of these skills and Appendix A for the full test papers.

[^11]:    ${ }^{18}$ Abhijeet Singh also describes these issues, as they relate to the economics of education in developing countries, in an excellent blog post here: https://blogs.worldbank.org/impactevaluations/how-standard-standard-deviation-cautionary-note-using-sds-compare-across-impact-evaluations, accessed November 1, 2021.
    ${ }^{19}$ Focusing on the US context, Kraft (2020) provides excellent guidance on how, and when, to use effect sizes for such comparisons.

[^12]:    ${ }^{20}$ In our sample, this eligibility criterion excluded roughly 13 percent of children at baseline who would otherwise be eligible according to our two remaining eligibility criteria: one, the child's age; and two, their not having previously attended school at grade 1 or higher.

[^13]:    ${ }^{21}$ Subtask 6 on both tests has no written component, making it somewhat different than all other subtasks, and less difficult in practice than other higher-level subtasks.

[^14]:    Does the child have any visible/noticeable disability? (circle as appropriate) No Yes (please specify)

[^15]:    Assessment start time:

[^16]:    Thank you, let's move to the next task.

[^17]:    Does the child have any visible/noticeable disability? (circle as appropriate)
    No Yes (please specify)

