

CHILD LABOR, INCOME SHOCKS, AND ACCESS TO CREDIT*

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Abstract

Although a growing theoretical literature points to credit constraints as an important source of inefficiently high child labor, little work has been done to assess its empirical relevance. Using panel data from Tanzania, we find that households respond to transitory income shocks by increasing child labor, but that the extent to which child labor is used as a buffer is lower when households have access to credit. These findings contribute to the empirical literature on the permanent income hypothesis by showing that credit-constrained households actively use child labor to smooth their income. Moreover, they highlight a potentially important determinant of child labor and, as a result, a mechanism that can be used to tackle it.

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1. Introduction

This paper examines the relationship between household income shocks, access to credit, and child labor. In particular, we investigate the extent to which income shocks lead to increases in child labor, and whether households with access to credit can mitigate the effects of these shocks.

These questions are important for three reasons. First, they point to a potentially important determinant of child labor and, as a result, at mechanisms that can be used to tackle it. This is particularly relevant since child labor is often viewed primarily as a consequence of poverty. For example, in a recent publication, the World Bank defined child labor as “one of the most devastating consequences of persistent poverty” (Fallon and Tzannatos [1998]). To some extent, the stylized facts bear out this view. In 1995, the incidence of child labor was 2.3 percent among countries in the upper quartile of GDP per capita, and 34 percent among countries in the lowest quartile of GDP per capita (see Dehejia and Gatti [2002] and Krueger [1996]). If poverty is the main cause of child labor, then the prevalence of child labor should decrease as countries develop. However, the relationship between poverty and child labor has been put into question by some recent within-country studies.¹ In particular, imperfections in labor markets, education and – the focus of this paper – credit markets might be at the root of child labor. In this case, policies designed to correct such imperfections directly might be an efficient means of tackling child labor.

Second, in the theoretical literature on child labor (reviewed below), lack of access to credit plays a central role in determining the prevalence of child labor. From the household’s point of view, child labor entails a trade-off between immediate benefits (increased current income) and, to the extent it interferes with the accumulation of the child’s human capital, potential long-run costs

¹ In their review of several studies, Canagarajah and Nielsen [1999] do not find clear evidence that poverty is associated with higher incidence of child labor within countries. In their study of Ghana, Boozer and Suri [2001] do not find a link between poverty and child labor, concluding that policies to alleviate poverty may not be appropriate to reduce child labor. Using data from Ghana and Pakistan, Bhalotra and Heady [2001] provide evidence that children of land-rich households are *more* likely to work (and less likely to be in school) than their counterparts in land-poor households. They label this finding the “wealth paradox” since it challenges the notion that child labor is observed

(lower future earnings potential). A number of recent models show that, by preventing households from trading off resources intertemporally in an optimal way, credit constraints give rise to inefficiently high child labor.²

Two key economic variables that influence households' demand for and ability to smooth resources inter-temporally are income shocks and access to credit. Despite their theoretical centrality (for example, see the discussion in Grootaert and Patrinos [1999]), little empirical research has been undertaken to examine the connection between income shocks, access to credit, and child labor. Indeed, to our knowledge, this is among the first studies explicitly examining this link.

Third, our work relates to an important literature on the permanent income hypothesis, income smoothing, and credit constraints. In particular, our work provides a bridge between studies such as Zeldes [1989], who documents the importance of credit constraints in preventing optimal smoothing of consumption over time, and Townsend [1994], who demonstrates that household consumption follows a smoother path than household income. As documented extensively in work by Morduch [1994, 1995, 1999], if households succeed in smoothing their consumption profile, but are credit constrained, they are likely to resort to mechanisms other than borrowing to cope with income shocks. This paper examines one such mechanism, namely child labor. In this context, our work is complementary to Jacoby and Skoufias [1997] and Kochar [1999], who examine respectively the role of schooling and parental labor supply as buffers to income shocks.

more often among children in poor households. A combination of labor market failures and ill-functioning land markets can explain this result.

² In outlining child labor issues and directions for the World Bank, Fallon and Tzannatos [1998] point out that child labor can have additional costs in terms of harmful effects on physical health and mental well-being (such as psychological and social adjustment) of children. Moreover, others contend that the working conditions for children are far below those of adults in terms of hours worked, wages, and safety.

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Using four rounds of household panel data from the Kagera region of Tanzania, we show that transitory income shocks – as proxied by crop accidentally lost to insects, rodents, and fire – lead to significantly increased child labor. Moreover, we find that households with collateralizable assets – which, in line with the existing literature (see for example Jacoby [1994]) we interpret as a measure of access to credit – are better able to offset the effects of these shocks. It is of course possible that a higher level of collateralizable assets mitigates the effect of child labor for reasons other than the ability to borrow. One of these confounding effects is a wealth effect and another possibility is that the level of collateralizable assets is correlated with unobservables such as a household’s social network. Nonetheless, we find that our results remain robust after controlling for other sources of wealth and for household fixed effects.

The paper is organized as follows. In Section 2, we provide a brief review of the literature. The empirical strategy is outlined in Section 3. In Section 4, we describe our data. In Section 5 we present our results and Section 6 concludes.

2. Literature Review

2.1 Theoretical Perspectives on Child Labor

Basu [1999] partitions the child labor literature into two groups: papers which examine intra-household bargaining (between parents, or parents and children) and those which examine extra-household bargaining (where the household is a single unit and bargains with employers).

In the intra-household bargaining framework, child labor is the outcome of an optimization process which places different weights on household members, for example parents and children (see Bourguignon and Chiappori [1994] and Moehling [1995]), or the mother (who altruistically cares for the children) and the father (who cares for himself in addition to the family; see Galasso [1999]). The weight that each member receives can depend upon his or her contribution to the family’s resources. Collectively, child labor may be beneficial because it contributes to family

income, and it may be desirable to the child because it increases his or her weight in the family's decision function. Within this framework, key variables in determining child labor are those that influence the relative bargaining strength of different members of the household. This could include wealth, the number, age, and gender of children, and individual wages as well as the set of opportunities for education available to children such as access to and quality of schooling.

The inter-household bargaining framework considers each household as a unitary entity (see Becker [1964] and Gupta [2000]). The motivation behind this approach is that children's bargaining power is inherently very limited. The parents and the employer bargain about the child's wage and the fraction of that wage to be paid as food to the child. Within this framework, the key variables are those that determine the relative bargaining strength of the household vis à vis the employer. These include, amongst others, household wealth and variables such as the access to credit.⁴

The unitary model of the family à la Becker [1964] is best suited to understanding the role of borrowing constraints as determinants of child labor. Analytically, the question is closely related to the bequests literature, which has highlighted that the non-negativity constraints in bequests can lead to an inefficient allocation of resources within the family (see for example Becker and Murphy [1988]). A recent strand of theoretical literature has addressed the implications of this for child labor (Parsons and Goldin [1989], Baland and Robinson [2000], and Ranjan [2001]). In particular, these papers show that if parents care about their children, but parents' bequests are at a corner, child labor is generally not efficient. The basic intuition is that child labor creates a trade-off between current and future income (Baland and Robinson [2000]). Putting children to work raises current income, but by interfering with children's human capital development, it reduces future income. The future income is, of course, realized by the children

⁴ Most theoretical models of child labor (and certainly the empirical literature) focus on supply-side factors. Canagarajah and Nielsen [1999] outline demand-side factors which can play critical roles in perpetuating child labor.

and not the parents. Thus, if bequests are positive, parents can compensate themselves for foregone current income by reducing bequests. Instead, when parents do not leave bequests for their children (for example, if they are poor) or financial market imperfections do not allow parents to trade off old-age income with current resources, parents will have their children supply “too much” labor.

This theoretical approach suggests that the availability of credit should be a factor that predicts the incidence of child labor. Moreover, we can interpret evidence of such an effect as an indication that observed child labor is inefficiently high.

2.2 *Empirical Work on Child Labor*

Although the recent theoretical literature highlights income shocks and borrowing constraints as an important source of inefficiency in the allocation of resources within the family and, in particular, of inefficiently high child labor (Ranjan [2001] and Baland and Robinson [2000]), the link between income shocks, access to credit, and child labor remains largely unexplored in the empirical literature.

There are a wide range of empirical studies of child labor at the micro-level using household survey datasets. These studies tend to estimate reduced-form participation equations for child work. For example, in a recent volume, Grootaert and Patrinos [1999] review findings of studies of child labor from Côte D’Ivoire, Colombia, Bolivia, and the Philippines. Canagarajah and Nielsen [1999] review findings from five studies from Côte D’Ivoire, Ghana, and Zambia. These are only several of a much wider list of studies in this area which have contributed to a better understanding of the factors that influence child labor. A consistent finding of these studies is that the child’s age and gender, education and employment of the parents, and rural versus urban

These include the non-pecuniary characteristics of children that may make them desirable employees for employers, such as being more willing to take orders and do monotonous work, and less aware of rights, among other factors.

residency are robust predictors of child labor. Few studies examine child labor in Tanzania using household survey data (see Beegle [1998]), although several studies focus on schooling determinants in Tanzania.⁵

There are a number of studies that assess the relationship between schooling outcomes and shocks. Of course, it is not obvious that for children there is a one-to-one trade-off between time spent in school and time spent working. Many children not enrolled in school are not necessarily working and, in many cases, enrolled children combine their schooling with work. Furthermore, hours in either activity may be sufficiently low on average that an increase in time spent in one activity will not crowd out time spent in another, as opposed to crowding out leisure time (for example, see the evidence in Ravallion and Wodon [2000]).

Jacoby [1994] examines the relationship between borrowing constraints and progression through school among Peruvian children. He concludes that lack of access to credit perpetuates poverty because children in households with borrowing constraints begin withdrawing from school earlier than those with access to credit. Jacoby and Skoufias [1997] is the first empirical work that rigorously addresses the issue that poor households lacking access to credit markets might draw upon child labor when faced with negative income shocks. Using data on school attendance patterns from six Indian villages, the authors conclude that fluctuations in school attendance are used by households as a form of self-insurance. However, these studies focus only on schooling and not on child labor activities, which would likely be more directly affected by income shocks and inability to access credit.

Dehejia and Gatti [2002] use cross-country data on child labor to investigate the association between child labor and access to credit (as proxied by financial development) across countries.

⁵ These studies examine a range of schooling outcomes, such as: the general decision to enroll at primary or secondary levels (Al-Samarrai and Peasgood [1998] and Mason and Khandker [1997]); delayed enrollment decisions (Bommier and Lambert [2000]); urban-rural differentials (Al-Samarrai and Reilly [2000]); the impact of own health and illness among household member on attendance (Burke [1998] and Burke and Beegle [2003]); and the relationship between orphanhood status and enrollment (Ainsworth et al [2002]).

They find a significant negative and robust relationship between the two variables, which is particularly sizeable in poor countries. Moreover, their evidence suggests that in the absence of developed financial markets, households resort substantially to child labor in order to cope with aggregate income variability.

The only work that we are aware of that uses micro data to examine the link between credit constraints and child labor is Edmonds [2002] and Guarcello et al. [2002]. Edmonds [2002] examines the impact of the South African old-age pension on the child labor of households that are headed by senior citizens. He, like us, finds evidence of credit constraints, but using a different strategy. We discuss his work again in the next section. Guarcello et al. [2002] analyze the response of child labor to broadly defined income shocks (loss of employment, death in the family, droughts in the region, etc.) in Guatemala. They identify a sub-sample of credit-constrained households using information on denial of credit for households that applied and self-reported information on why a family was not able to apply for credit among non-applicants. They find that credit rationing is associated with higher child labor and that shocks significantly increase the proportion of working children. The results in Guarcello et al. [2002] are complementary to ours although they do not evaluate the differential impact of shocks in households facing different levels of credit constraints. While the authors have the advantage of availing themselves of direct information on access to credit, in their setup lack of access to credit might reflect household characteristics associated with attitudes towards child labor rather than financial market imperfections. Similarly, the cross-sectional structure of their data does not allow them to fully disentangle the extent to which shocks are exogenous.

2.3 *The Permanent Income Hypothesis Literature*

As indicated in the introduction, this paper bridges two strands of the empirical literature on the permanent income hypothesis (PIH).

One strand of this literature examines whether credit constraints can be used to account for apparent rejections of the PIH. In particular, Zeldes [1989] splits households in his sample by their holding of financial assets and examines the extent to which the Euler equation implied by the PIH hold for both groups. He concludes that credit constraints are an important impediment to intertemporal consumption smoothing for many households. A second strand of the literature examines the extent to which households can insure themselves against idiosyncratic income shocks. Studies showing that the profile of household consumption is smoother than the profile of household income indicate that households are able to some extent to smooth away income shocks (see Deaton [1992], Morduch [1995], and Townsend [1994]). The two results together suggest that households – even when credit constrained – are resorting to some means other than borrowing (or explicit insurance) to smooth away shocks.

In this paper we examine whether child labor is one such mechanism that can account for household income smoothing. In particular, we examine the response of agricultural households to a transitory income shock. To the extent that the shock is transitory, the PIH suggests that households should borrow to smooth away much of the shock. In this context, our test has two parts. First, do households with limited borrowing capacity indeed increase child labor in response to a transitory shock? And, second, is this effect mitigated for households with greater borrowing capacity? As such, our work is in line with Paxson [1992], who documents that households save most of their transitory income and Jacoby and Skoufias [1997], who demonstrate households adjust their children's school attendance in response to shocks.

Edmonds [2002] uses a different, albeit related, identification strategy. He examines the impact of a permanent increase in old-age pension support in South Africa. Whereas our

identification of credit constraints is based on finding an effect of a transitory income shock, Edmonds examines a fully anticipated increase in income. If households are not credit constrained, then they can borrow against the future receipt of their old-age pension, and there should be no adjustment in child labor at the time the household actually begins to receive its pension. The two approaches are clearly complementary. A caveat to Edmonds' approach is that households may not borrow against their future old-age pension simply because they are myopic, not credit constrained. In this case, the effect he estimates would simply be a wealth effect. An advantage of our approach is that households do not have to be forward-looking to respond to a negative income shock. Of course, our key underlying assumption is that the income shock that we examine is indeed transitory.

3. Empirical Strategy and Specification

In this section, we outline the empirical strategy and specification we employ to identify the effects of income shocks and access to credit on child labor choices within households. We are interested both in the direct effect of an income shock as well as in the extent to which access to credit helps families to smooth away the impact of a shock.

The literature on consumption and savings distinguishes between two types of shocks – transitory and permanent. The theoretical effect of these shocks on time allocation and borrowing decisions are different. If we believe that income affects child labor outcomes negatively (with a stronger relationship among the poor), we would expect a permanent negative income shock to increase child labor, especially for children in poor households. Furthermore, we do not expect households to borrow to offset the effect of such shocks. Conversely, credit or self-insurance through accumulated savings can be an effective tool in smoothing transitory shocks. Instead, increasing child labor in response to a transitory shock carries important costs for human capital development because of the often-irreversible disruption in schooling.

In this paper, we focus on transitory shocks, as proxied by the value of crop loss due to insects, rodents, fire and other calamities (we describe the characteristics of these shocks in more detail in the next section). We interpret finding a significant effect of shocks on child labor decisions as evidence of incomplete credit (and insurance) markets at the household level. In this context, we would then expect increased access to credit to offset the effect of transitory shocks. The policy implications of this result are relevant – improving household access to credit can mitigate the effect of income shocks and thereby reduce inefficiently high child labor.

Measuring credit constraints is particularly difficult. Ideally, one would want an indicator identifying those households that sought credit but could not obtain it, as well as a measure of how much credit they wanted. In practice, this is rarely observed. The literature suggests, though, that access to credit is correlated with collateral (Townsend [1994] and Jacoby [1994]). In this paper, we observe the value of collateralizable assets (this is discussed in greater detail in the following section).

We first examine the effect of shocks on child labor hours. Our basic specification is:

$$(1) \quad y_{ijt} = \beta_0 + \beta_1 X_{ijt} + \beta_2 shock_{ijt} + \varepsilon_{ijt}$$

where subscripts index individuals (i), households (j), and survey rounds ($t=1, \dots, T$); y is child labor hours, $shock$ is our measure of the income shock (discussed in detail in the next section), and X contains a set of controls including individual, household, and community characteristics. We expect transitory shocks to lead to an increase in child labor if credit or insurance are limited or non-existent, i.e. we expect $\beta_2 > 0$.

We then investigate the role of access to credit. In particular, we estimate the following specification:

$$(2) \quad y_{ijt} = \beta_0 + \beta_1 X_{ijt} + \beta_2 shock_{ijt} + \beta_3 (shock_{ijt} \cdot collateral_{ijt}) + \beta_4 collateral_{ijt} + \varepsilon_{ijt}$$

Here the effect of interest is β_3 , which captures the differential impact of a shock among households with different levels of collateralizable assets. We expect access to credit to mitigate the effect of transitory shocks, i.e. $\beta_3 < 0$.

We first estimate these equations with OLS, where we pool all rounds of our panel survey together and allow for clustering at the household level. Even though this specification controls for a wide range of observable individual and household characteristics and includes community dummies, there are many potential dimensions of selection along unobservable characteristics. For example, less forward-looking households might be both more prone to income shocks (for example, because they are less careful in the management of their fields) and, at the same time, more inclined to sending their children to work (because they place less value on formal education).

To address the problem of selection on unobservables, we allow for fixed effects and estimate the following:

$$(1') \quad y_{ijt} = \alpha_j + \delta_t + \gamma_w + \beta_1 X_{ijt} + \beta_2 shock_{ijt} + \varepsilon_{ijt}$$

$$(2') \quad y_{ijt} = \alpha_j + \delta_t + \gamma_w + \beta_1 X_{ijt} + \beta_2 shock_{ijt} + \beta_3 (shock_{ijt} \cdot collateral_{ijt}) + \beta_4 collateral_{ijt} + \varepsilon_{ijt}$$

where α_j , δ_t , γ_w , are household, time (season), and survey round fixed effects respectively. As a result, the model is estimated using within-household variation controlling for seasonal and survey

round effects. Note that fixed effects at the household level subsume community-level unobservable characteristics. However, we cannot rule out time-varying household-level unobservable characteristics.

4. Data Description and Summary Statistics

The data for this study are from a panel dataset for the Kagera region in Tanzania. The Kagera Health and Development Survey (KHDS) was part of a research project conducted by the World Bank and the University of Dar es Salaam. The KHDS surveyed over 800 households in the region up to four times from 1991-1994 with an average interval between surveys of six to seven months.⁶ Households are drawn from 51 communities, mostly villages, in the six districts of Kagera.

This dataset has several features that make it particularly appropriate for the proposed analysis. First, the detailed household survey has a wide array of individual and household characteristics, including information on time use of all household members aged seven and older. This includes time spent in the previous week working on household businesses (farm and non-farm), for wages in non-household business, and in household chores. The household survey also includes information on crop loss, as well as measures of physical and financial assets in each of the four interviews. The data are longitudinal and, as such, they allow us to test and control for unobservable variables that may bias cross-sectional results.

Our definition of child labor is the total hours in the last week spent working in economic activities and chores (including fetching water and firewood, preparing meals, and cleaning the house). Economic activities for children consist predominately of farming, including tending crops in the field, processing crops, and tending livestock. Few children are engaged in wage employment or non-farm family businesses. This is consistent with reports of limited child-labor

⁶ The explicit objectives of the KHDS were to measure the economic impact of fatal illness (primarily due to HIV/AIDS) in the region and to propose cost-effective strategies to help survivors. For more information about this project, see Ainsworth et al. [1992] and World Bank [1993].

markets from the community data. We include chores as well as economic activities for two reasons. First, the concept of child labor (by ILO standards) is not restricted to only economic activities.⁷ Second, in the largely rural sample of households in this study, it may be difficult to distinguish time in household chore activities and time spent preparing subsistence food crops. For our study, we focus on two age samples. Our primary age group includes children 10-15 years old. We also include a second group, children 7-15 years old. Given the well-documented low enrollment levels and delayed enrollment in Tanzania, along with the low hours of work among younger children (7-9), our main focus are children of age 10 and older. The upper age range for child-labor studies is typically 14 or 15 years, the age of completed primary schooling if enrolled on time.

Table 1 presents summary statistics of the sample in our study, broken down by the three samples on which the regressions are run. In addition, the last 2 columns show summary statistics for the main sample separated into those children in households that had experienced an income shock and those that had not. In the pooled data, children worked on average about 21 hours in the previous week. Mean hours as well as most other covariates are similarly distributed in households with and without a shock. More than 90 percent of children worked at least 1 hour in the last week. About one-third of children reside in households that report some crop loss.⁸ About one-half of children reside in households with any durables, our primary indicator of collateral. Among those households that experienced a shock, the total value of the shock is about twice (in log terms) the value of per capita durable goods. The prevalence of alternative wealth indicators, cash holdings and physical assets, is larger. Nearly three-quarters of children live in households with some cash and all children live in households with some physical assets (including the value of land, business

⁷It should also be mentioned that the concept of child labor does not necessarily refer to simply any work done by a child, but, rather, work that stunts or limits the child's development or puts the child at risk. However, in survey data it is difficult (perhaps impossible) to appropriately isolate the portion of time spent working on the farm that qualifies under this very nuanced definition. Therefore, we follow the standard convention in the empirical literature.

⁸ For zero values of shock, durables, cash, and physical assets, the bottom value is coded at 1.

equipment, livestock, and dwellings). The average household size is quite large, over 7 members on average. In part, this reflects the sample of households with children; in the entire sample (including households with no children age 10-15), the average size is about 5.7 members. Levels of parental education are extremely low. Few children have fathers who had attended school beyond the primary level (12 percent of children) and almost none had mothers with more than primary schooling (2 percent).

Turning to our measure of income shock, for the identification strategy to be credible, we ideally want an income shock that is: of a sufficient magnitude to potentially affect household time allocation; exogenous to child labor decisions; and transitory. The data include reports of the value of crop loss due to insects, rodents, and other calamities (such as fire) in each survey round. We compute the total value of crop loss for all crops farmed. This measure of income shock has several advantages. First, since agriculture is the main economic activity in the Kagera region, many households experience a shock and these shocks are extremely relevant with respect to household income. As reported in Table 2, Panels A and B, 88 percent of households experience a shock during one of the four survey rounds, and around 50 percent of those households experiencing one shock lose 25 to 75 percent of the value of their crops to calamities. Hence these shocks are common, and can be significant in magnitude.

Second, these shocks are plausibly exogenous and transitory. Table 3, Panel A, shows the correlation between the occurrence of a shock and several household characteristics. As discussed in Morduch [1994], in general there may be reasons to believe that poorer households (with a high prevalence of child labor) may also be more prone to income shocks. However, none of the standard household covariates predicts shocks in our data. Moreover, controlling for household characteristics, shocks do not appear to be correlated over time.

One might argue that households can use information not available through the survey to forecast shocks. If that is the case and families actively use child labor as a buffer against shocks,

they might adjust child labor supply in advance. However, we find that child labor in period $t-1$ does not significantly predict shocks (for value as well as share of crop loss) in period t (Table 3, Panel B). Overall, this evidence suggests that the value of crop loss provides us with credible transitory, exogenous, and unanticipated shocks.

We measure access to credit with collateralizable assets. This is in keeping with the related literature, including Zeldes [1989] and Jacoby [1994]. Consistent with this approach, fully 80 percent of all lenders (banks, NGOs, private individuals) reported requiring collateral for loans.⁹ As noted by Jacoby, it is important to select a class of assets that is associated with a household's borrowing capacity, but is not directly linked to its demand for child labor. Collateralizable assets in this setting include the value of durable goods, such as radios, bicycles, fans, lamps and pots. This excludes cash holdings, business, and land value which might be directly correlated with demand for child labor within the household. In the regressions, we measure collateral as the (log) value of durable assets per household member.

5. Results

We first discuss the effect of income shocks on child labor as estimated with OLS and fixed effects. We then present estimates for the interacted model with shocks and collateral. Finally, we conduct several robustness checks, including using alternative definitions for shocks and access to credit, and running the specification on a suitably restricted sample.

5.1 *The Effect of Shocks*

Table 4, column 1, reports OLS estimates of the effect of the income shock on child labor for children between the age of 10 and 15. An income shock is associated with significantly higher child labor. A one standard deviation income shock is associated with a 10 percent increase in

mean hours in the last week. This effect is robust in magnitude and significance to including community dummies in the regression (column 2).

Column 3 reports the estimates from a household fixed effect model. Since we are now controlling for a substantial range of variables (including effects related to households, seasons, and survey rounds), the estimated coefficient is smaller than the OLS estimate (0.22 vs. 0.33) but still statistically significant at the 1 percent level. This result confirms that the correlation between shocks and child labor is not driven by spurious unobserved family effects. We obtain similar results if we expand our sample to the 7-15 age range (column 4). Not surprisingly, we find that the magnitude of the coefficient on shocks decreases once the sample is enlarged to include 7-10 year old children. First, younger children work less on average. Moreover, we expect their work supply to be less responsive to shocks – to the extent that parents perceive them as less productive than their older siblings, they will be more willing to send the older children to work in response to a shock. We interpret all of these findings as evidence that households resort to child labor substantially when coping with income shocks.

5.2 *The Effect of Credit*

In Table 5, we estimate equations (2) and (2') using OLS (column 1) and fixed effects (column 2 for 10-15 year olds, column 3 for 7-15 year olds). These models allow for both income shocks and access to credit, and in particular, for a differential response of child labor hours to shocks in households with different levels of collateralizable assets. Columns 4-7 extend the model to include additional wealth controls and restrict the sample to households with land.

First, note that the effect of the income shock is stable across columns 1-3 and similar in magnitude and significance to the estimates reported in Table 4.

⁹ The community questionnaires provided this information.

The OLS estimates (column 1) return a negative and significant coefficient on collateral, but no significant effect for the interaction between collateral and shock. As discussed in Section 3, OLS estimates are potentially biased due to selection effects. There are two sources of concern. First, households with higher levels of collateral may also tend to have lower levels of child labor, simply due to wealth effects. Second, unobservable factors such as motivation and inter-temporal discount factors might lead some households to value the long-term benefits of education as well as to undertake actions to avoid income shocks. In both cases the between-household association of access to credit and child labor would be non-causal. Fixed effects in column 2 allow us to focus on the within-household relationship between collateral and income shocks. There we see that the estimated impact of collateral is completely absorbed by fixed effects – in particular, it switches sign and becomes insignificant. Conversely, the interaction between shock and collateral is consistently negative and significant in the fixed effect estimates. This suggests that the availability of credit mitigates the effect of shocks. In particular, a one standard deviation increase in the income shock is associated with a 10 percent increase in child labor hours for households without durable assets, whereas in a household with a median level of durables, the increase in child labor hours is 5 percent. In other words, the availability of a typical level of collateral offsets half the effect of a shock.

Estimates are similar in significance and magnitude for the sample of children between age 7 and 15 (column 3) and are in line with results in Table 4, column 2.

It could be argued that the value of durables simply captures a wealth effect as opposed to the effect of access to credit. In order to preclude this alternative explanation, in columns 4-6 we control for other measures of wealth, namely cash holdings and per capita physical assets (land, equipment, and tools). Under these alternative specifications, the interaction of interest remains significant and stable, while the direct effect of the additional wealth controls is not significant.

To confirm the validity of our estimated effects, we examine the subset of households that is most directly affected by agricultural shocks. We do so in two ways. First, we include an indicator for whether the head of the household's main activity is farming. This indicator is not significant and its inclusion does not affect our estimates of interest (column 5). However, this indicator is not always a precise measure of the extent to which farming contributes to family income. As an alternative, we use the size of land ownership as an indicator of whether farming is likely to be the main income source for the whole household. We confine ourselves to the sample of households with more than one acre of land (anecdotally, this is generally considered the minimum land size for subsistence) – and also exclude the top 1 percent of landowners. The results are virtually unchanged.

It is also possible that these additional sources of wealth matter differentially when the household is faced with an income shock. In particular, we would expect that households tap their cash holdings first when faced with a shock but resort to mortgaging land – for which there exist only thin markets in the region – only as an extreme remedy to a transitory shock. As such, cash can in principle perform a similar function to collateralizable durables, while ownership of physical assets should not, on average, matter in the interaction with the shock. It is also possible that wealth variables proxy for the extent of households' social networks, which arguably are more valuable in times of need. When we include these three sources of wealth and their interaction with the shock in the same specification and estimate it with fixed effects (column 7), income shocks remain strongly and positively associated with child labor. However, these additional wealth variables are highly collinear, and as a result the statistical significance of our coefficient of interest diminishes, although its magnitude is robust.

5.3 *Robustness Checks*

In Table 6, we examine the sensitivity of our results to the definition of income shock and collateral. One concern with our definition of income shock is that we might be more likely to observe large crop losses among richer households – the more you have, the more you would expect to lose if you have a loss. However, an indicator of whether a significant crop loss occurred would still identify households without necessarily inducing a strong correlation with the level of income. When we use this alternative definition of shock (adjusted to encompass income shocks of some relevance, above 10 percent of the total value of crop), we find that, similar to Table 5, shocks are positively and significantly associated with child labor hours (Table 6, column 1). Crop losses of 10 percent or more of overall crop value are associated with a 7 percent increase in child labor hours. Moreover, in column 2 we find that the interaction between this new measure of shock and collateral is negative and significant. Children in households with no durables have 12 percent higher labor hours (evaluated at mean hours) when crop losses occur, compared to 4 percent higher hours for children in households with median per capita durable assets. Column 3 reports the results for the sample restricted by land ownership. The increase in child labor hours associated with income shocks widens between children in households with and without collateralizable assets (8 percent and 19 percent, respectively).

We then examine alternative definitions of collateral while returning to our initial measure of income shock. Because of the large share of zeros in the distribution of durables (somewhat less than 50 percent), it is reasonable to examine the robustness of our result to a specification that uses an indicator for zero versus positive levels of collateral. In column (4), we note that the estimated coefficient on the interaction of interest is unchanged in sign and significance, although of course the magnitude changes since collateral is now a dummy rather than a continuous variable. One could also argue that it is the total value of durables in the household, as opposed to its per capita value, that matters as collateral. To account for this possibility, we use the total durable value as an

alternative measure of collateral in column 5. When we do so, the estimates of our effects of interest are virtually unchanged, even while controlling for cash holding and value of physical assets (column 5).

6. Conclusions

This paper has examined the link between transitory income shocks, access to credit, and child labor. We show that child labor increases significantly in response to crop losses due to calamities such as fire, insects, rodents, etc. Moreover, we show that the availability of collateralizable assets offsets the effect of income shocks on child labor, even when controlling for other sources of wealth in the family and for household level unobservables. This result is important because it corroborates a large theoretical literature on the relevance of credit constraints in predicting child labor, points to child labor as one of the mechanisms that households use to smooth transitory income shocks, and finally suggests that expanding access to credit might be effective in mitigating the prevalence of child labor.

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Table 1: Summary Statistics

	(1)	(2)	(3)	(4)	(5)
	Ages 10-15	Ages 7-15	Ages 10-15 Land restriction	Ages 10-15 with shock	Ages 10-15 without shock
Hours:					
Mean	21.23 (14.91)	18.13 (14.83)	21.27 (14.87)	22.41 (16.35)	20.62 (14.09)
% > 0	0.94	0.90	0.94	0.95	0.94
log value of crop loss:					
Mean	2.92 (4.21)	2.96 (4.23)	3.00 (4.26)	8.60 (1.87)	0.00
% > 0	0.34	0.34	0.35	1.00	0.00
log per capita durables:					
Mean	4.28 (4.13)	4.21 (4.12)	4.30 (4.08)	4.22 (4.03)	4.31 (4.18)
% > 0	0.54	0.53	0.55	0.55	0.53
log per capita cash:					
Mean	5.24 (3.21)	5.20 (3.20)	5.25 (3.16)	5.11 (3.29)	5.30 (3.16)
% > 0	0.78	0.78	0.79	0.75	0.80
log per capita physical assets:					
Mean	11.20 (1.44)	11.15 (1.46)	11.28 (1.28)	11.15 (1.35)	11.23 (1.48)
% > 0	1.00	1.00	1.00	1.00	1.00
Farm household =1 if yes, else 0	0.76 (0.43)	0.76 (0.43)	0.79 (0.41)	0.77 (0.42)	0.76 (0.43)
Household size	7.86 (3.74)	7.83 (3.66)	7.99 (3.77)	7.83 (3.41)	7.87 (3.90)
Father's schooling: 1-6 years =1 if yes, else 0	0.43 (0.50)	0.42 (0.49)	0.46 (0.50)	0.45 (0.50)	0.42 (0.49)
Father's schooling: 7 years =1 if yes, else 0	0.30 (0.46)	0.33 (0.47)	0.30 (0.46)	0.28 (0.45)	0.31 (0.46)
Father's schooling: 8+ years =1 if yes, else 0	0.12 (0.32)	0.12 (0.32)	0.10 (0.29)	0.13 (0.34)	0.11 (0.32)
Mother's schooling: 1-6 years =1 if yes, else 0	0.37 (0.48)	0.35 (0.48)	0.37 (0.48)	0.39 (0.49)	0.35 (0.48)
Mother's schooling: 7 years =1 if yes, else 0	0.28 (0.45)	0.31 (0.46)	0.26 (0.44)	0.28 (0.45)	0.29 (0.45)
Mother's schooling: 8+ years =1 if yes, else 0	0.02 (0.13)	0.02 (0.13)	0.01 (0.09)	0.02 (0.13)	0.02 (0.13)
Observations	3839	5591	3234	1302	2537

Notes: Standard deviations are in parentheses.

Table 2: Frequency and Magnitude of Shocks

Panel A: Frequency of shocks

Number of shocks across four survey rounds	Number of households	%
0	58	12.0
1	224	46.4
2	166	34.4
3	33	6.8
4	2	0.4
Total	483	100

Panel B: Magnitude of shocks, among households with one shock

Share of the value of crop loss to total crop value	%
1-24% (low)	38.8
25-49% (medium-low)	24.1
50-74% (medium-high)	25.0
75-100% (high)	12.1
Observations	224

Table 3: Predicting the Occurrence of Shocks

Panel A: Any crop loss		
	(1)	(2)
Specification:	Probit	Probit
Lagged any crop loss		-0.030 (0.084)
Head of the HH years of schooling	0.028 (0.018)	0.025 (0.018)
Head of the HH is female	0.187 (0.119)	0.205* (0.120)
Head of the HH's age	0.001 (0.003)	0.001 (0.003)
log per capita durables	0.008 (0.010)	0.008 (0.011)
log per capita cash	-0.027 (0.017)	-0.027 (0.017)
log per capita physical assets	0.031 (0.037)	0.031 (0.035)
Observations	1698	1698
R-squared	0.30	0.32

Notes: Standard errors are in parentheses. *** indicates significance at 1%; ** at 5%; and, * at 10%. Community, district, survey round, and season dummies are included but not reported.

Panel B: Crop loss in period t conditional on child labor in period t-1

	(1)	(2)
Specification:	FE	FE
Dependent variable:	log value of crop loss	share lost
Mean child labor hours (t-1)	-0.003 (0.007)	0.0001 (0.000)
Head of the HH years of schooling	-0.070 (0.156)	-0.014* (0.008)
Head of the HH is female	-1.237* (0.745)	-0.062 (0.039)
Head of the HH's age	0.033 (0.023)	0.001 (0.001)
log per capita durables	0.002 (0.053)	0.001 (0.003)
log per capita cash	-0.035 (0.034)	-0.003* (0.002)
log per capita physical assets	0.030 (0.098)	0.001 (0.005)
Number of observations	1755	1755
Number of households	668	668
R-squared	0.33	0.25

Notes: Standard errors are in parentheses. *** indicates significance at 1%; ** at 5%; and, * at 10%. Dependent variables are log value of crop loss and share of crop loss due to shock to total crop value. Survey round, and season dummies are included but not reported.

Table 4: Hours Worked in the Last Week and Income Shocks

	(1)	(2)	(3)	(4)
Specification:	OLS	OLS with	FE	FE
	Ages 10-15	community dummies	Ages 10-15	Ages 7-15
		Ages 10-15		
Shock: log value of crop loss	0.33*** (0.09)	0.30*** (0.09)	0.22*** (0.08)	0.17*** (0.07)
Father's schooling: 1-6 years	-1.61* (0.94)	-1.31 (0.93)	-0.36 (1.61)	-0.55 (1.20)
Father's schooling: 7 years	-1.64* (0.98)	-1.69* (0.98)	0.13 (1.61)	0.44 (1.22)
Father's schooling: 8+ years	-1.44 (1.34)	-0.61 (1.27)	-2.20 (2.26)	-0.14 (1.63)
Mother's schooling: 1-6 years	-0.61 (0.79)	-0.76 (0.74)	0.44 (1.34)	0.15 (1.01)
Mother's schooling: 7 years	-0.02 (0.89)	0.27 (0.83)	1.54 (1.40)	0.43 (1.00)
Mother's schooling: 8+ years	-7.36*** (2.01)	-3.72** (1.83)	1.82 (3.38)	-0.23 (2.55)
Observations	3839	3839	3839	5591
Number of households	636	636	636	716
R-squared	0.05	0.09	0.04	0.15

Notes: Standard errors are in parentheses. In columns 1 and 2, standard errors are computed correcting for heteroskedasticity and correlation within household clusters. Fixed effects are computed at the household level. Hours include time spent on economic activities and household chores. *** indicates significance at 1%; ** at 5%; and, * at 10%. Other regressors included, but omitted from the table, are age and age squared and indicator variables for missing parental education, the season at time of interview, district (columns 1 & 2) and the round of the interview.

Table 5: Hours Worked in the Last Week, Income Shocks, and Collateral

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Specification:	OLS	FE	FE	FE	FE	FE	FE
	Ages	Ages	Ages	Ages	Ages	Ages	Ages
	10-15	10-15	7-15	10-15	10-15	10-15	10-15
						Land restriction	Land restriction
Shock: log value of crop loss	0.34*** (0.11)	0.34*** (0.10)	0.27*** (0.08)	0.33*** (0.11)	0.33*** (0.11)	0.37*** (0.11)	1.01* (0.58)
Collateral: log per capita durables	-0.18** (0.085)	0.13 (0.15)	0.06 (0.12)	0.12 (0.15)	0.12 (0.15)	0.20 (0.17)	0.18 (0.17)
Shock * collateral	-0.01 (0.02)	-0.03* (0.01)	-0.02** (0.01)	-0.03* (0.01)	-0.03* (0.01)	-0.04** (0.02)	-0.02 (0.02)
Liquid wealth: log per capita cash				-0.13 (0.10)	-0.13 (0.10)	-0.18 (0.11)	-0.14 (0.14)
Shock * Liquid wealth							-0.01 (0.02)
Illiquid wealth: log per capita physical assets				0.47 (0.30)	0.47 (0.30)	0.61 (0.38)	0.77 (0.41)
Shock * Illiquid wealth							-0.06 (0.05)
Farm household					-0.59 (1.00)		
Father's schooling: 1-6 years	-0.91 (0.93)	-0.35 (1.61)	-0.57 (1.20)	-0.35 (1.61)	-0.35 (1.61)	-0.77 (1.75)	-0.78 (1.75)
Father's schooling: 7 years	-1.39 (0.96)	0.070 (1.61)	0.42 (1.22)	0.016 (1.61)	0.010 (1.61)	-0.051 (1.80)	-0.081 (1.80)
Father's schooling: 8+ years	-0.35 (1.26)	-2.30 (2.26)	-0.17 (1.63)	-2.34 (2.26)	-2.33 (2.26)	-2.99 (2.65)	-3.00 (2.65)
Mother's schooling: 1-6 years	-0.71 (0.74)	0.40 (1.34)	0.13 (1.01)	0.40 (1.34)	0.39 (1.34)	0.05 (1.41)	0.05 (1.41)
Mother's schooling: 7 years	0.45 (0.82)	1.53 (1.40)	0.41 (1.00)	1.49 (1.40)	1.48 (1.40)	1.35 (1.49)	1.34 (1.49)
Mother's schooling: 8+ years	-2.87 (1.91)	1.84 (3.38)	-0.22 (2.55)	1.85 (3.38)	1.81 (3.38)	-1.74 (4.20)	-1.69 (4.20)
Observations	3839	3839	5591	3839	3839	3234	3234
Number of households	636	636	716	636	636	571	517
R-squared	0.10	0.04	0.15	0.04	0.04	0.04	0.04

Notes: Standard errors are in parentheses. In column 1, standard errors are computed correcting for heteroskedasticity and correlation within household clusters. Fixed effects are computed at the household level. Hours include time spent on economic activities and household chores. *** indicates significance at 1%; ** at 5%; and, * at 10%. Column 1 includes community dummy variables. Other regressors included, but omitted from the table, are age and age squared and indicator variables for missing parental education, the season at time of interview, district (column 1) and the round of the interview.

**Table 6: Hours Worked in the Last Week, Income Shocks, and Collateral:
Robustness Checks**

	(1)	(2)	(3)	(4)	(5)
Specification:	FE	FE	FE	FE	FE
	Ages 10-15	Ages 10-15	Ages 10-15	Ages 10-15	Ages 10-15
			Land restriction	Land restriction	Land restriction
Definition of shock:	Dummy for crop share lost>.1	Dummy for crop share lost>.1	Dummy for crop share lost>.1	log(crop loss)	log(crop loss)
Definition of collateral:		Per capita durables	Per capita durables	Dummy for durables>0	Total durables
Shock	1.48** (0.75)	2.47*** (0.96)	3.38*** (1.06)	0.34*** (0.12)	0.36*** (0.12)
Collateral		0.10 (0.15)	0.18 (0.17)	1.42 (1.25)	0.15 (0.14)
Collateral * shock		-0.28** (0.14)	-0.33** (0.16)	-0.24* (0.13)	-0.03** (0.01)
log per capita cash		-0.13 (0.10)	-0.18 (0.11)	-0.18 (0.12)	-0.16 (0.11)
log per capita physical		0.51* (0.30)	0.64 (0.38)	0.61 (0.38)	0.43 (0.38)
Household size					-0.83 (0.22)
Father's schooling: 1-6 years	-0.35 (1.61)	-0.35 (1.61)	-0.78 (1.75)	-0.76 (1.75)	-1.13 (1.74)
Father's schooling: 7 years	0.12 (1.61)	-0.01 (1.61)	-0.05 (1.80)	-0.01 (1.80)	(1.80)
Father's schooling: 8+ years	-2.19 (2.26)	-2.33 (2.26)	-2.96 (2.65)	-2.95 (2.65)	-3.48 (2.65)
Mother's schooling: 1-6 years	0.45 (1.34)	0.40 (1.33)	0.06 (1.41)	0.071 (1.41)	0.15 (1.41)
Mother's schooling: 7 years	1.58 (1.40)	1.53 (1.40)	1.36 (1.49)	1.35 (1.49)	1.46 (1.49)
Mother's schooling: 8+ years	2.00 (3.38)	1.85 (3.38)	-1.89 (4.20)	-1.85 (4.20)	-1.86 (4.19)
Observations	3839	3839	3234	3234	3234
Number of households	636	636	517	517	517
R-squared	0.04	0.04	0.04	0.04	0.05

Notes: Standard errors are in parentheses. Fixed effects are computed at the household level. Hours include time spent on economic activities and household chores. *** indicates significance at 1%; ** at 5%; and, * at 10%. Other regressors included, but omitted from the table, are age and age squared and indicator variables for missing parental education, the season at time of interview, and the round of the interview.