

Health and “Poverty Traps” in Rural Developing Economies

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Abstract

This paper tests for the presence of a positive convex relationship between individual consumption and productivity using data from rural Ethiopia. Such a convexity underlies the theory of efficiency wages and can create a low-consumption, low-productivity (“poverty”) “trap” operating through health status. I propose a test based on the predictions of a dynamic theoretical model of household consumption allocation which obviates the need to directly estimate the relationship between individual consumption and productivity directly. This is particularly advantageous, as at least one of these variables is typically not observed for individuals. Using semi-parametric estimation techniques, I find evidence of a convexity in the consumption-productivity relationship for households in the upper part of the land-per-worker distribution. Productivity incentives and the preference for equality appear to be in conflict for these rural households, who therefore face the possibility of a low-health, low-productivity steady state. Results also show that it is important to use a broad measure of an individual’s health or work capacity, as using a standard nutrition-based measure (body mass index) obscures these dynamics.

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1. Introduction

Ever since Leibenstein (1957) and Mirrlees (1975) advanced the notion of a link between individuals' consumption and productivity, there has been interest in studying this link. In developing countries, interest has focused particularly on the prospect of dual causality between nutrition and income. "Nutrition"- or "efficiency- wage" models (e.g., Stiglitz [1976], Dasgupta and Ray [1986, 1987] and Ray and Streufert [1993]) posit a convexity in this relationship in the low income and nutrition range. That is, for some range of nutritional status the marginal productivity of consumption (or nutrition) is increasing in consumption. If such a convexity is present, then a low nutrition-low productivity ("vicious") circle can produce an equilibrium with persistently under-employed and under-nourished workers (Ray and Streufert [1993]). Yet a focus purely on nutrition may fail to fully capture the essential linkages between consumption and productivity. Strauss and Thomas (1998) note that the nature of work in low income economies tends to rely more heavily on strength and endurance and, therefore, on good health generally. The body's store of calories is only one component of this. Yet in part due to a paucity of data, little research has been done using a broader measure of health status.

Empirical studies focusing on nutritional status have found some support for a causal linkage between nutrition and wages (see e.g. Berhman and Deolaker [1989], Foster and Rosenzweig [1994], and Strauss [1986]) but on the whole cast doubt on the nutrition-efficiency models (Binswanger and Rosenzweig [1984], Rosenzweig [1988]; for a review see Strauss and Thomas [1998]). This could be at least in part because maintaining recommended caloric intakes is cheap relative to wages, even in some poor economies such as rural India (Subramian and Deaton [1996] and Swamy [1998]). Nonetheless, in many countries nutritional sufficiency is a perennial issue. Moreover, the convexity between

consumption and work capacity posited by efficiency “wage” models may become more realistic when consumption relates to all health inputs rather than merely to calories.

Consumption-productivity models are, of course, premised on the presence of dual causality between income and health, and this makes the relationships difficult to identify empirically. In particular, there are simultaneity issues not only between health and income, but also with other important production inputs (see Smith [1999] for a survey of the health-income literature). Moreover, income is measured with a great deal of error in agricultural households. More importantly, often either individual income or individual consumption data are not available. For these and other reasons, it has been extremely difficult to identify dual causality in general, and in particular whether there is a convexity present in the relationship.

This paper tests for both of these phenomena using data from rural Ethiopian households. It uses a broader measure of health which is likely to be more closely related to productivity or work capacity than is nutritional status. An indirect estimation strategy is proposed which avoids many of the difficulties of estimating the consumption-productivity relationship directly, particularly as neither of these are observed for individuals.

While it is often supposed that health affects income in poor rural areas, one can imagine circumstances in which this is not the case, at least in the short run. For example, if household members are under-employed due to a lack of productive assets (and general unemployment), then in the event one member falls ill, his or her work could be done by another. Nonetheless, strenuous work is part of the life of most rural Ethiopians; thus health is likely to affect income even in the short run. Income is also likely to affect health, as it can help a household to command a healthier lifestyle – vegetables and fruits, meat and other

protein, warm clothing, soap, energy for boiling water, and medical care.¹ At the same time, if access to quality medical care is not affected by income (or if, as in some parts of rural Ethiopia, it is not available at all) then a poor health shock may persist similarly for relatively rich and poor households. Alternatively, if a household member contracts a chronic illness, effective medical treatments may simply not be available at any price. Finally, if there is no market for health inputs such as clean water or “clean” (non-toxic) energy, then there may be little the better-off household can do to obtain these inputs. Nonetheless, it is clear at least in the extreme that income and wealth affect health, as it is the poorest and landless who have most suffered the persistent ill effects of famine (including death.) (See Webb *et al.* [1992] for an analysis of the effects of famine in Ethiopia).

In this paper, I develop a dynamic intra-household allocation model which takes dual causality as given. This model yields testable implications which would be unlikely to hold in the absence of such causation. Moreover, it admits a test for the presence of a productivity tradeoff with the households’ desire to equalize health outcomes across its members – that is, a convexity in the consumption-productivity relationship. In particular, I estimate movements in the relative health status of pairs of individuals within the same household with respect to their initial health status. A finding that initial health status affects the degree of equalization (or dispersion) in the pair’s subsequent health would support a general model in which households weigh equality and productivity considerations in their intra-household resource allocation. If initial health enters negatively in some range, then – as the theory

¹Despite a burgeoning literature for the developed world, the exact mechanisms by which income “buys” better health are still in great need of study. There are also likely to be direct intra-household spillover effects from a poor income or health draw. If household members are too weak to carry fresh water in sufficient quantities, for instance, to take children for health checks, or to collect heating fuel, then other members of the household may suffer ill health.

shows – this is evidence that productivity incentives dominate for some range, and thus that a convexity is present.

The results I obtain support the presence of dual causality. Households are generally able to improve the health of their members as incomes improve. Thus even in areas, like rural Ethiopia, with extremely poor health care provision it appears that resources matter. In addition, I find evidence that households in the upper part of the distribution of landholdings (per worker) confront an important productivity-equality tradeoff. This implies that health also affects income (*i.e.*, productivity). It also implies a convexity in the relationship between consumption and productivity. Thus these households can neither maximize their productivity nor fully invest in the recovery of the less healthy. Moreover, this convexity can be a mechanism for multiple equilibria (“poverty trap”) in which low productivity and poor health persist for some time.²

The next section presents the theoretical model and the propositions which form the basis of the empirical tests. Section 3 describes the data and the necessary institutional background. Section 4 describes the empirical strategy, and Section 5 reports the results. Section 6 concludes.

2. Theoretical Model

I develop a model in which uncertainty is explicit with respect to both health status and income, as such uncertainty is an important feature of poor rural economies. I begin with a health transition equation, which relates health of individual i in period t to health in period $t+1$:

$$h_{t+1}^i = f(g_t^i + f^{-1}(h_t^i(1-\delta))) + \varepsilon_{t+1}^i \equiv f(\tilde{g}) + \varepsilon_{t+1}^i, \quad (1)$$

where g_t^i represents current consumption of all goods and services – food, warm clothing, clean water, shelter, leisure, etc. – by the individual which contribute to health (“health goods”). \tilde{g} represents the (depreciated) “stock” of health goods stored in the body in the prior period plus their current consumption. h_t denotes health status in period t , δ is the systematic (natural) rate of deterioration of health from one period to the next with no further g investment. ε_{t+1}^i is a pure random shock to the health status of individual i which is unrelated to his consumption and is centered at zero. Here I assume that the $f(\tilde{g})$ function does not vary by individual, but this assumption could be relaxed without invalidating the approach. I assume that $f'(\tilde{g}) > 0$ everywhere, as is commonly assumed for poor populations. According to (1), health status is somewhat cumulative, but one can raise it through greater consumption, and maintenance of health requires some ongoing investment. It may not always be possible for physiological and other reasons to improve (chronically) poor health through expenditure of resources, and in this case $f(\tilde{g})$ would contain a flat segment.

In addition to (1), there is a household income production function in which the lagged health status of all members of household H ($1, 2, \dots, N^H$) enters:

$$y_{t+1}^H = y^H(h_t^1, h_t^2, \dots, h_t^{N^H}; L, \Gamma) + \eta_{t+1}^H \quad (2)$$

where H indexes the household and N^H denotes the number of working age members of household H . L denotes the household’s landholdings. These can be taken as exogenous given the lack of a legal market in rural land in Ethiopia. All land was redistributed by the Revolutionary (*Derg*) government in 1975, with only slight adjustments since (with a

²A longer panel than that available here would be needed to pin down the long run dynamics

correlation of .99 between land reportedly held following this redistribution and that held today). η_{t+1}^H represents a random shock to income which is not related to lagged health status but could be correlated with ε_{t+1} according to the joint distribution function $F(\eta, \varepsilon)$. One can also write (2) from an individual productivity perspective, given other members' health (h_t^{-i}) as follows:

$$y_{t+1}^i = y^i(h_t^i | h_t^{-i}; L, \Gamma) + \eta_{t+1}^i \quad (3)$$

The exact shapes of functions (1) and (3) will of course depend upon the metric used for health. Nonetheless, it is typically assumed that the functions are concave in their upper ranges. That is, when health status is very good, it becomes more difficult to improve through greater g consumption, and there is a natural upper bound for good health. Moreover, it is reasonable to assume that health status exhibits decreasing returns at some point. However, it is not clear *a priori* whether there will be a region of convexity at lower values of h_t^i .³

Thus there are two general alternatives for the shapes of $y(\tilde{g})$ and $f(\tilde{g})$, which are familiar from Mirrlees (1975) and Dasgupta and Ray (D&R) (1986). Figure 1 represents Hypothesis A, that $f(\tilde{g})$ and $y(\tilde{g})$ are concave throughout the range relevant to the population. Under this scenario, even at poor health states one may easily improve health through a more diverse diet, simple hygienic practices, or basic medical interventions. Moreover, individual production as a function of g consumption is concave due to the concavity of either or both of (1) and (3). Hypothesis B is represented in Figure 2. According to this hypothesis, it requires relatively more investment to bring a very low health individual

empirically or indeed to estimate the structural model with any confidence.

³Although each individual's function (1) will vary somewhat due to genetic or historical factors, the basic shapes will depend upon common human physiology, as well as the price and availability of

to the point where he can perform physical tasks. More formally, if one or both of (1) and (3) is sufficiently convex in some range, then the composite function, $y_{t+1}^H(\tilde{g}_{t-1}^i) \equiv y(\tilde{g})$, which relates i 's g consumption in period t to income in $t+2$, would exhibit an S-shape, as shown.

D&R (1986) show that if such an S-shaped relationship (convexity) exists between nutrition and productivity, then a vicious circle can arise wherein individuals remain unemployed and undernourished. There are some important differences between their framework and that used here, however. First, in the D&R model individuals can work for an outside (equilibrium) wage as long as they are sufficiently productive, whereas here production occurs at the household level. Thus a consumption-productivity circle could arise for the individual only if he is employed (rather than unemployed) in the sense that the household normally assigns him productive tasks.⁴ Being unable to perform such tasks would lower household income. On the other hand, the marginal product of an individual household member could be close to zero if the ratio of land (and other capital) to people is low and if there are few outside income generation possibilities. In this case, a single individual's poor health status may have no effect on subsequent income. Moreover, if the $f(\tilde{g})$ curve is sufficiently steep, a temporarily ill individual may still be fed, clothed, and cared for by the household and his or her illness may not affect household income beyond one period.

In the dynamic version of the D&R model, Ray and Streufert (1993) have shown that some individuals will remain malnourished and involuntarily unemployed indefinitely, even

health goods, and thus should be somewhat similar across individuals within the economy. Moreover, one can control for differences in (3) through individual, household, and village characteristics.

⁴ I will not consider the outside labor option formally, though this does represent a small portion of some households' income.

with no missing markets.⁵ However, their model contains no uncertainty, a critical feature of the environment considered here. Under uncertainty within a household-production system, the likelihood of multiple steady states even in the presence of an S-shaped consumption-productivity relationship is much less clear.

I assume that household utility is separable in individual utility and over time and that the household H has the following optimization problem:

$$\max_{c_{t+1}^i, g_{t+1}^i \text{ all } i} E_t \sum_{l=0}^{\infty} \beta^l \sum_{i=1}^{N_H} \theta^i u(c_{t+l}^i, g_{t+l}^i, h_{t+l}^i), \quad (4)$$

where θ^i represents individual i 's pareto weight in household resource allocation and $\sum_{i \in H} \theta^i = 1$. The variable g represents health goods and c represents consumption of goods which do not affect health (such as, for example, coffee, spices, decorative items, or a second pair of shoes).⁶ I ignore for simplicity the existence of goods that may damage health, such as cigarettes, alcohol, and other non-medicinal drugs. In the empirical strategy, I am able to take an agnostic view on which goods are to be considered c goods versus g goods, as I do not use consumption data directly.

Household resources are denoted by Z_t and follow the transition equation below, where the safe return to liquid assets is denoted r :

$$Z_{t+1}^H = (Z_t^H - \sum_{i=1}^{N^H} g_t^i - \sum_{i=1}^{N^H} c_t^i)(1+r) + y_{t+1}^H \quad (5)$$

Carroll (1997) and Osborne (2002) show that in theory liquidity constraints do not matter appreciably for households' long run (steady state) consumption behavior. Risk

⁵ The ability to borrow would break the vicious circle in partial equilibrium with identical individuals, but not in general (dynamic) equilibrium where the labor market is explicitly modeled and individuals differ in their landholdings.

aversion alone will make perfect consumption smoothing over time sub-optimal.⁷ In general, households will behave differently when they are at a low versus a high asset state relative to their long run mean wealth, whether they can borrow or not. Thus, there is no compelling reason to assume that households cannot borrow. A standard transversality condition can ensure that all borrowings are repaid (or indefinitely serviced).

I assume that the utility function is concave in all its arguments and that households strongly prefer to avoid catastrophic health (or death) of members:

$$\begin{aligned} u_k &> 0 \quad \forall k = g, c, h \\ u_{kk} &< 0; \\ u_h(g, c, h) &\rightarrow \infty \text{ as } h \rightarrow 0 \end{aligned} \tag{6}$$

Under standard regularity assumptions, one can write down a Bellman Equation as follows:

$$V(Z_t, h_t^1, h_t^2, \dots, h_t^{N^H}) = \max_{g_t^i, c_t^i \forall i} \sum_{i=1}^{N^H} \theta^i u(g_t^i, c_t^i, h_t^i) + \beta E_t V(Z_{t+1}, h_{t+1}^1, h_{t+1}^2, \dots, h_{t+1}^{N^H}) \tag{7}$$

The First Order Conditions for g_t^i (for all i) are:

$$\theta^i \frac{\partial u}{\partial g_t^i} + \beta \frac{\partial f}{\partial \tilde{g}_t^i} E_t \left[\frac{\partial V_{t+1}}{\partial h_{t+1}^i} \right] = \beta(1+r) E_t \left[\frac{\partial V_{t+1}}{\partial Z_{t+1}} \right] \tag{8}$$

This says that the pareto-weighted instantaneous marginal utility of g consumption plus the discounted expected marginal value of future health (the investment value of g) equals the discounted expected value of lost resources (Z) from the purchase of g .

Equating these conditions across all individuals $i \neq j$ in the household results in:

⁶ These examples assume that the psychic (or stress-reducing) effects of possession or consumption of such items does not affect health appreciably.

⁷Households who know that they cannot borrow are considered “constrained,” even if in the current period they do not wish to borrow.

$$\theta^i \frac{\partial u}{\partial g_t^i} + \beta \frac{\partial f}{\partial \tilde{g}_t^i} E_t \left[\frac{\partial V_{t+1}}{\partial h_{t+1}^i} \right] = \theta^j \frac{\partial u}{\partial g_t^j} + \beta \frac{\partial f}{\partial \tilde{g}_t^j} E_t \left[\frac{\partial V_{t+1}}{\partial h_{t+1}^j} \right] \quad (9)$$

The current marginal utility of consumption of g plus the discounted expected future marginal value of health status multiplied by the “price” of marginal health will be equal across individuals in the household. Noting also the following for all $i \in H$:

$$\frac{\partial V_t}{\partial h_t^i} = \theta^i \frac{\partial u}{\partial h_t^i} + E_t \left[\beta \frac{\partial V_{t+1}}{\partial Z_{t+1}} \frac{\partial y_{t+1}}{\partial h_t^i} + \beta(1-\delta) \frac{\partial V_{t+1}}{\partial h_{t+1}^i} \right], \quad (10)$$

we see that the marginal value of individual i 's health is higher the more productive is his future health status and the higher is his marginal utility of (future) health – under Hypothesis A, that is, the lower is his future health.

To develop testable implications I must assume that allocative preferences within the household are determined only by productivity and/or equality concerns, and therefore that $\theta^i = \theta^j$ for all $i, j \in H$. This is a strong assumption. However, one can deal with the possible bias caused by unobserved pareto weights by controlling for observed characteristics, as discussed below.

Given the theory presented above, one can draw upon results from similar models in the precautionary savings and consumption literature (Deaton [1991], Carroll [1997]) to argue that under Hypothesis A the household will have a preference for relative equality of health outcomes:

Proposition 1:

If (1) and (3) are strictly increasing and concave (Hypothesis A) and $\theta^i = \theta^{i-1}$ for all $i, i-1$ such that $h_t^i > h_t^{i-1}$ households will seek greater equality in h_{t+1}^i and h_{t+1}^{i-1} such that :

$$E_t \left[\frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}} \right] \equiv E_t R_{t+1} < 1.$$

Proof. See Appendix 1.

Moreover, the strength of the equalizing incentive will increase the lower is initial health:

Proposition 2:

Under the Assumptions of Proposition 1, the degree of equalization will be decreasing as health status improves in the sense that $E_t \left[\frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}} \right]$ is rising in h_t^i given the initial health dispersion, $h_t^i - h_t^{i-1}$.

Proof. See Appendix 1.

The household will weigh each individual's expected (marginal) contribution to

future resources through the term, $\beta^2 E_{t+k-1} \frac{\partial V_{t+k+1}}{\partial Z_{t+k+1}} \frac{\partial y_{t+k+1}}{\partial h_{t+k}^i} \frac{\partial h_{t+k}^i}{\partial \tilde{g}_{t+k-1}^i}$, which equals

$\beta^2 E_{t+k-1} \frac{\partial V_{t+k+1}}{\partial Z_{t+k+1}} y'(\tilde{g}_{t+k-1}^i)$. If the $y(\tilde{g})$ curve is everywhere concave, then this will only

reinforce the equalizing incentive.⁸ Under Hypothesis A, therefore, $E[R]$ is lower when health status is low (and $y(\tilde{g})$ relatively more concave). In addition, under this hypothesis relative equalization incentives will be greater (and $E[R]$ lower) when h_t^i and h_t^{i-1} are relatively far apart.

If the assumptions of the Propositions are not true, however – in particular, if there is a convexity in $y(\tilde{g})$, the household may prefer to equalize health outcomes less. A

⁸ Allowing for individually-varying curves, $f^i(\tilde{g}^i)$ and $y(h^i | h^{-i})$, one can generalize this framework in terms of a household-level marginal productivity curve, denoted $y'(f^{-1}(h(1-\delta)))$ where each member of the household is arrayed along the horizontal axis in the order of his/her initial health status

“productivity effect” may produce a non-monotonic or negative relationship between $E[R]$ and h .⁹ This scenario (Hypothesis (B)) is illustrated in Figure 2. Here households will allocate somewhat less to individual $i-1$ and more to individual i than would otherwise be the case. These two individuals could even see their health status diverge in expectation so that $E_t R_{t+1} > 1$, whereas individuals j and $j-1$ would tend to see theirs converge.

Equation (10) also shows that the productivity effect is “weighted” by the expected marginal value of the household’s future resources, which will be high when resources are relatively low. Thus households with low resources relative to their long run mean will need to weight productivity considerations more heavily.

Dercon and Krishnan (2000) (D& K) use these same data to test for “perfect” sharing of the risk of (unpredicted) illness within the household. They use a static model similar to that used in the broader risk sharing literature (e.g., Chiappori 1992, Townsend 1994). The model used here differs from theirs in that health goods consumption enters the utility function directly, and this creates a source of persistence in intra-household inequality. In addition, uncertainty is explicitly accounted for in the dynamic optimality conditions, since households may wish (depending upon r and δ) to “save” more in lower health individuals due to risk aversion. In the presence of these two factors pareto optimality (as in Equation (9)) looks somewhat different than the standard model implies. Moreover, in deriving their test, D&K do not allow for different “user costs” of improving nutritional status across

(or more precisely $f^{-1}(h(1-\delta))$). If this curve exhibits an upward sloping portion or inverted U-shape, then equalizing and productivity incentives come into conflict.

⁹ If only one of (1) or (3) exhibits a convexity, but $y(\tilde{g})$ does not, then there is no productivity tradeoff, and households will seek to equalize health of their members.

members over time, which here are represented by $\frac{\partial f}{\partial g}$ and $\frac{\partial y}{\partial h}$. Testing for this type of potential divergence is at the core of this exercise.¹⁰

The model as specified is fairly general, and without selecting parameters and functional forms it can produce a variety of dynamic behaviors. Moreover, without solving it numerically for given parameters and functional forms, the long run stationary distribution(s) for health and income cannot be known. However, it is clear that for some combinations of $F(\varepsilon_t, \eta_t)$, $f(\tilde{g})$, and $y(h)$, multiple steady states (*i.e.*, poverty traps) may arise. In a similar, stochastic growth model with (capital) stock-dependent utility by Kurz (1968) and Nyarko and Olson (1991, 1994), it is impossible to rule out *a priori* multiple steady states, even with strictly concave functional forms. On the other hand, Nyarko and Olson (1994) show that under further restrictions to the utility function and (their model's analogue) to $y(\tilde{g})$, if there is sufficient uncertainty, there will only be one limiting distribution – the preferred steady state.¹¹ Ruling out Hypothesis A in this model does not necessarily rule out multiple steady states, just as rejecting it does not necessarily imply such multiplicity. However, given that there is a great deal of uncertainty in the context of rural Ethiopian agriculture, a multiplicity of steady states is particularly unlikely under Hypothesis A. The strategy used here, therefore, is to use Propositions 1 and 2 to test the null of an everywhere concave consumption-productivity relationship (Hypothesis A).

¹⁰ It is easy to show that under their model with CRRA utility, with no productivity issue in the sense of different $\frac{\partial y}{\partial h}$ or $\frac{\partial f}{\partial \tilde{g}}$ across individuals, even if pareto weights are unequal, households will attempt to change all individuals' health status by the same percentage. This would imply a constant R with respect to h_t^i .

3. The Data and the Measurement of Health

The data used are from the Ethiopian Rural Household Survey (1994-1995) (ERHS), conducted by the Addis Ababa University and the Centre for the Study of African Economies at Oxford University. There are 1476 households included in the survey and approximately 9,820 individuals. There are three rounds of data, collected in the first instance from May-August 1994, in the second from January-March 1995, and finally from August-November 1995. Households were asked about the health of all individuals in the household in all three rounds. There are generally two production cycles in the Ethiopian agricultural calendar – a large main (*meher*) harvest, and small (*belg*) harvest. In the first round, households had recently experienced (and reported on) a *meher* harvest. There had been another *meher* harvest for some (6 out of 15) villages by Round 2, and by Round 3 for all villages. In addition, there was a *belg* harvest between Rounds 1 and 2 for most villages. I will assume that between each round households had an income draw, decided their consumption levels ($g_t^i \forall i$), and experienced a new shock to their health, so that what one observes in the survey is the resulting health status, $(h_{t+2}^i(\tilde{g}_t^i))$. In terms of the model, that is, (at least) two periods transpire between rounds.

Measuring health status is a non-trivial issue, as health measures are often of necessity either categorical in nature or limited to measures of nutritional status (*i.e.*, body measurements). The ERHS contains each individual's weight and height in each round, so that a reasonable short run measure of nutritional status, the body mass (or Quetelet) index

¹¹The model used here differs from theirs in that investment in health itself contributes to instantaneous utility.

can be calculated as weight in kilograms over height in meters squared. This has been found to be a reliable (cardinal) measure of nutritional status.¹²

The ERHS also contains data which capture various measures of health and work capacity. In each round an informed household member was asked how easily each member over the age of 7 could perform the following tasks: Stand up after sitting down; sweep the floor; walk for 5 kilometers; carry 20 liters of water for 20 meters; and hoe a field for a morning. I assigned a score of 4 to the answer of “easily,” 3 for “with a little difficulty”, 2 for “with a lot of difficulty” and 1 for “not at all.” Although these measures are somewhat subjective, asking respondents to focus on specific tasks may make their answers less subjective than a measure ranging, for example, from “poor” to “excellent” health. Moreover, they are reasonable measures of work capacity and thus well suited to the issues examined here. My point of departure is to assume that they contain some information on health status which is not contained in the body mass index (*bmi*). Indeed, there is no clear mapping from them to *bmi* in these data.¹³

As one might expect, the scores fall with the difficulty of the task, so that more people had difficulty hoeing than, say, standing up.¹⁴ In addition, men were systematically more able to perform all tasks, including those normally considered a woman’s task (such as carrying water or sweeping). Health scores for women were, moreover, much more variable than for men. Table 1 shows the means and standard deviations by sex of each individual task-measure. All pregnant or breast-feeding women were excluded from the sample. In addition, only “working age” adults from the ages of 16 to 64 are included. This is to avoid

¹²See James, Ferro-Luzzi, and Waterlow (1988).

¹³Unlike for the other capacity measures, for instance, *bmi* is decreasing in the score for the ability to stand up.

¹⁴To reduce the problem of missing data, if either walk, carry, or hoe were missing and the individual could not stand up after sitting down or sweep the floor, I assumed that they could not carry water or

the issues surrounding child physical development, which are somewhat distinct from health status. Moreover, elderly individuals (> 64) may respond less predictably to health investments and are thus excluded.

As there is no clearly superior way to aggregate these individual task measures into a single health measure, I begin by simply adding them together (and subtracting 4) to create a health “index” (*hlthindx*) (on a 1-16 point scale). Thus impaired ability to stand up is rather arbitrarily counted the same as impaired ability to hoe a field, and increasing impairment is counted the same as milder impairment in performing more tasks. Nonetheless, if one cannot stand up, one is not likely to be able to hoe a field either, and if one has severe impairment in a less difficult task, one is likely to be severely impaired in more difficult tasks as well. The survey responses are consistent in these respects. Thus, the aggregate measure (*hlthindx*) should not be a bad measure of the true ranking of individuals’ health status. However, it will blur distinctions in health status of individuals who have the same health score.

Note that health scores were on average significantly worse in the first round of the survey. *Hlthindx* averaged 15.03 in Round 1 versus 15.32 and 15.38 in Rounds 2 and 3, respectively. Round 1 corresponded to the “lean” season, when cash inflows are typically lower and when plowing and preparing the land for the following main harvest are also done. Moreover, the 1994 lean season was a relatively difficult one, as both the previous *meher* and contemporaneous *belg* harvests were poor and food prices had risen dramatically. This appears to have had an important effect on health status, an effect which is to some extent reversed over the subsequent rounds.

Using this index as a measure of health status poses several difficulties. First, it is highly skewed. Table 2 (Column 2) shows the distribution of *hlthindx* in the pooled (3-

hoe either. This mainly affected the “hoe” variable, as there were few examples of missing values for

round) sample. Over 80 percent of the sample has the top score of 16, and there are very few observations in the lower part of the distribution. This makes inference somewhat difficult. Nevertheless, deviations from 16 are likely to be quite meaningful, as are all differences between health scores.

The second major issue in using *hlthindx* is its non-cardinality. Thus one cannot, for example, compare the difference in true health status between the scores 2 and 3 with that between 14 and 15. Thus one would like to transform this measure into a cardinal (continuous) one. To do so I assume that there is an underlying latent measure of health, h^* which is generated by the following process:

$$h_r^{i*} = W_r^i \pi + u_r^i,$$

where W is a vector of determinants of health status, π is a vector of coefficients, and u is an error term distributed as a standard normal. Observations are indexed by individual (i) and by round (r). The observed, or reported, health measure is generated as

$$hlthindx = m(h^*),$$

where m is a function mapping the latent measure into the ordinal outcomes $j, j=1 \dots J$. I wish to obtain the conditional expectation of h^* given the reported value (*hlthindx*) and the W vector. This is given as

$$E[h_r^{i*} | hlthindx_r^i, W_r^i] = W_r^i \hat{\pi} + E[u_r^i | W_r^i, hlthindx_r^i], \quad (11)$$

where $W \hat{\pi}$ is the contribution due to the observables, $\hat{\pi}$ is the ordered probit estimate of π , and $E[u | \cdot]$ is the component due to unobservables. An estimate of this unobservable component is known as the generalized residual (see Vella, 1993) and is constructed as:

“carry”.

$$\sum_{j=1}^J \delta_j^i \frac{\phi(\hat{\mu}_{j-1} - W_r^i \hat{\pi}) - \phi(\hat{\mu}_j - W_r^i \hat{\pi})}{\Phi(\hat{\mu}_{j-1} - W_r^i \hat{\pi}) - \Phi(\hat{\mu}_j - W_r^i \hat{\pi})} \quad (12)$$

where the $\hat{\mu}$ represent the estimated cut points and $\delta_j^i = 1$ if individual i is in category j of $hlthindx$ and 0 otherwise. Φ and ϕ represent the cumulative distribution function and the density function of the standard normal distribution. (More details on the construction of $hlthhat$ are presented in Appendix 2.) $hlthhat$ can only be estimated up to scale and position, but the variable of interest, $R = \frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}}$ as defined in *Proposition 1* is invariant to linear transformations of the health status measure. Thus after estimation I also transform the measures linearly to make them more comparable to the original $hlthindx$.

Table 2 (Column 3) shows the distribution for $hlthhat$ for the full sample (pooled rounds 1-3, for working age people.) This measure shows much less skewness than $hlthindx$ (Column 2), and as a continuous, cardinal measure of “true” health, avoids many of the problems inherent in using $hlthindx$.¹⁵ The sample used in estimation is displayed in Table 2 (Column 4).¹⁶

Before proceeding to formal testing, it bears noting that the overall trend over the three rounds is towards improvement in health status, as well as greater equalization across the sampled population. To see this, I conducted a simple descriptive OLS regression of the change in health on initial health as follows:

¹⁵There are other problems as well with using $hlthindx$. Frequently more than one individual in a household has the same health measure. Thus in pairing individuals, more than one may be matched to the same $i-1$ individual, and observations would not be strictly independent. In addition, a possible selectivity issue would arise in that individuals in households with no dispersion in $hlthindx$ in a given round would have to be dropped from the analysis, as there is no way to pair them with a less healthy individual (though this problem is somewhat mitigated by the fact that these are almost uniquely cases where $h=16$).

¹⁶ These are those for which data are complete, for rounds 1 and 2 only, and after trimming (see below).

$$\Delta h_{r+1}^i = \alpha + \lambda h_r^i + \epsilon_{r+1}^i \quad (13)$$

where r indexes round, $r=1,2$, and where ϵ_{r+1}^i is a random error term. The negative (and statistically significant) coefficients on initial health of $-.54$ and $-.34$ for *hlthindx* and *hlthhat*, respectively suggests that on average individuals who suffer from poor initial health, whether due to a pure shock or to household allocation decisions, will on average see greater improvements to their health over the medium run than those with better health. However, this simple correlation masks some of the underlying intra-household dynamics.

4. Empirical Strategy and Issues

The purpose of the empirical estimation is twofold: (1) To test whether the dynamic health distribution among household members is broadly consistent with the theory – that is, whether current health status affects the implicit allocation of g , all else equal; and (2) To test the null of Hypothesis A.

I define the convergence ratio R as in Proposition 1:

$$R_{r+1}^i = \frac{h_{r+1}^i - h_{r+1}^{i-1}}{h_r^i - h_r^{i-1}} \quad (14)$$

where h_{r+k}^{i-1} represents the $r+k$ round health status of the person in individual i 's household whose round r health was next-worst to his/hers among household members. This measure of convergence is convenient as household-level (linear) shocks to health do not affect it.¹⁷ I assume that the households' target ratio and realized R are related in the data through the following equation:

$$R_{r+1}^i \equiv E_r R_{r+1}^i(\tilde{X}_r^i) + \frac{\varepsilon_{r+1}^i - \varepsilon_{r+1}^{i-1}}{h_r^i - h_r^{i-1}} \equiv ER_{r+1}^i + e_{r+1}^i \quad (15)$$

where target R depends upon a set of conditioning variables \tilde{X} which may include h_r^i . e_{r+1}^i is a pure random error term of unknown distribution centered at zero. The average realization of R_{r+1}^i is thus a measure of the intensity of the household's incentives to equalize versus disperse target health of i and $i-1$ in $r+1$. Rewriting $h_r^i - h_r^{i-1}$ henceforth as *hlthdif*, the estimating equation can be written generally as:

$$\begin{aligned} R_{r+1}^i &= p(h_r^i) + X\beta + e_{r+1}^i \\ &\equiv p(h_r^i) + \gamma \cdot \text{hlthdif} + \Gamma^i x_r^i + \Gamma^{i-1} x_r^{i-1} + \Psi z_r^i + e_{r+1}^i \end{aligned} \quad (16)$$

where $X = (\text{hlthdif}, x, z)$, x is a vector of individual characteristics and z represents village-time effects and household characteristics. The x and z variables are included to control for observable determinants of health and productivity as well as households' constraints and opportunities. Since (as in *Proposition 2*) one must also control for the initial dispersion in health to test Hypothesis A, I include *hlthdif* in the equation. γ , Γ , and Ψ are unknown parameters, and $p(h)$ is some unknown function.

Given *Proposition 2*, if estimated $p(h)$ is increasing in some range this would be consistent with the primacy of equalization incentives. A decreasing slope in some range would, however, indicate the presence of a productivity tradeoff. Thus, one can test the null hypothesis (A) by testing whether or not there is a decreasing range of $\hat{p}(h)$. In general, this test is biased in favor of the null, since this tradeoff may still exist even if households'

¹⁷ Such shifts would affect a measures such as $\frac{h_{r+1}^i}{h_r^i} - \frac{h_{r+1}^{i-1}}{h_r^{i-1}}$, however. The latter measure also has the disadvantage that a 1-unit improvement in health is in effect measured as a greater improvement the

equalizing incentives dominate for the applicable range of h . The hypothesis that households do not weigh the factors considered in the model in deciding their short run health investments would be consistent with their being no effect of current health status on the household's incentives to equalize (or dis-equalize) the health of its members. That is,

$$\frac{\partial p(h_i)}{\partial h_i} = 0.$$

Since equalization and productivity effects may (under Hypothesis B) offset each other in the determination of R , I split the sample according to households' presumed intensity of individual (own) labor demand on the farm, measured as landholdings per working age person (denoted *Landpbig*). One might expect, given the paucity of outside employment opportunities in rural Ethiopia, that households with relatively more land per worker would utilize their own labor more intensively and would face stronger productivity incentives in the allocation of goods across members. At the same time, they may be better able to afford equality. Thus under Hypothesis A, they would seek to equalize health more than would low *Landpbig* households. Under the alternative hypothesis, however, they may show a greater tendency to dis-equalize. Therefore, all the estimations were done for "low" *Landpbig* (< .05 hectares per person), and a relatively "high" *Landpbig* (> .05 hectares per person) separately.¹⁸ As discussed above, land was largely allocated to households after the 1974 Revolution on the basis of political objectives, and landholdings are unlikely to be correlated with unobserved health status. *Landpbig* itself may be correlated with such status through the effect on fertility decisions made 16 years ago; however, these fertility decisions

lower is initial health. I do not wish to impose this assumption.

¹⁸ This split was done rather arbitrarily and no other splits were examined. Given the skewness of landholdings, this sample split did not result in equal proportions in each sample. Making a 50-50 split would have resulted in a threshold for *Landpbig* that seemed unreasonable low. Although the high landholdings group has a higher average health score (8.71 versus 8.50) its average R (1.11) is not statistically significantly different from that of the low landholdings group (1.10).

(and more recent in- and out- migration) are pre-determined as of the first round of the survey. As such, they should not be correlated with recent unexplained shocks or adjustments to health.

Clearly an important issue related to estimation of (16) is that of simultaneity between h_r^i , $hlthdif$, and R . In particular, R will tend to be negatively correlated with $\varepsilon_r^i - \varepsilon_r^{i-1}$, as households will wish to offset these unplanned (random) shocks in period $r+1$. In addition, if pareto weights across individuals are not equal (i.e., $\theta^i > \theta^{i-1}$), a source of bias in the estimation of $p(h)$ could arise. That is, an individual with a high pareto weight will tend to have a higher health status, and the household may also tend to see his status increase vis-à-vis that of individual $i-1$.

Given that the form of $p(h)$ is unknown, and that the variables h and $hlthdif$ are endogenous, I use the semi-parametric instrumental variables estimator proposed by Newey, Powell, and Vella (1999) (hereafter NPV). That is, one approximates the shape of $p(h)$ by polynomial series, and to account for the endogeneity of h and $hlthdif$ one includes in the main equation additional series approximations for the unknown functions $\omega(resid)$ using the estimated residuals ($resid$) from two corresponding first stage regressions. Here $resid=(res, resdif)$ for h_r^i and $hlthdif$, respectively. Although I use a linear term in $hlthdif$ to control for initial dispersion in health (*Proposition 2*), I will allow the endogeneity of this term to enter non-linearly. The exact functional forms for $\hat{p}(h)$ and $\hat{\omega}(resid)$ are chosen through the use of cross-validation (CV), where the sum of squared residuals is minimized from the leave-one-out out-of-sample predictions (see Pagan and Ullah, 1999). The series approximation to $p(h)$ is allowed to be of order 0 through 10, and for $\omega(resid)$ of order 0

through 3. Cross validation has been shown to provide the “best” estimate of the function in the sense that mean squared error is minimized asymptotically.

For identification one must exclude at least one exogenous variable in each of the first stage equations from the second stage (main) equation, and this excluded variable must not be correlated with the errors in (16). In the first stage particularly I focus on parsimony. Only variables which are significant, nearly significant, or which logically form part of a group which is jointly significant (such as village-round effects or age and age squared) are included in the equations. I select the included variables from a list of exogenous or predetermined variables on the basis of these criteria. I then test the appropriateness of excluding any of the selected variables by examining their individual significance in the second stage regression once $p(h)$ and $\omega(resid)$ have been estimated. If any previously excluded variable is marginally insignificant or significant, I update the specification to include it in the second step (and re-do CV to update the chosen series specifications).

Candidate exogenous variables to be included in all steps as appropriate are as follows: (1) variables capturing household-reported exogenous events during the prior main (*meher*) production cycle, such as the incidence of cold temperatures (*cold*), excessive rain (*toomuch*), flooding (*flood*), an index of problems with the timing and amount of rainfall (*rainprob*), and other weather-related occurrences; the incidence of pests (*pest*), or destruction of crop by livestock (*trampled*); (2) variables controlling for the asset position of the household as of Round 1 of the survey, including household assets (*totassets*), the value of livestock (*SalVal*), and *Landpbig*, again as a measure of both land wealth and the intensity of individual labor demand on the farm. Also available is a measure of the quality of the household’s landholdings (*i.e.*, the percentage of landholdings of highest quality (*Lem*, denoted *pcLem*)).

Candidate controls (x_r) for individual characteristics of both individual i and individual $i-1$ may also be included if significant. These are age, age squared, sex (*male*), whether both individuals are female (*bothfem*), and age and age squared interacted with *male*; whether each individual is mainly engaged in farm work (*farmer*), whether he or she is head of the family (*head*), whether he has a chronic condition (*chronic*), and height. The latter are pre-determined measures of long-run health status, and although not strictly exogenous are unlikely to be correlated with the errors in (16). Furthermore, it is important to control for the presence of a chronic condition in those equations where appropriate, as the cost of improving health can differ for those with such a condition.

Household and village characteristics that are candidates for inclusion (z) are village dummies interacted with round dummies to control for village level shocks to production, health, and prices; the educational level of the head of the household (*eduhead*), whether there is anyone pregnant or breastfeeding in the family (*pregghm*), whether it is a female-headed household (*femhead*), dummy variables for ethnicity, the number of teens and adults (*totbig*), the number of children (*totsmall*) in the household, and the number of deaths in the household due to illness in the past five years (*Illdth*).¹⁹

Implementation of the NPV estimator involves two additional issues – overfitting and trimming. First, as discussed in NPV (1999), theory requires overfitting so that the bias is smaller than the variance asymptotically. However, since I do not wish to introduce more fluctuations into $\hat{p}(h)$ than are confirmed by CV, inference will be done using the model chosen by CV, although I will show both the CV and the overfitted estimates.

¹⁹Fertility decisions are taken as pre-determined, although it may be the case that healthier couples have more children.

Second, NPV argue that trimming may be required to avoid issues associated with outliers. Unfortunately, the trimming procedure one uses is unavoidably arbitrary. My approach is therefore to use reasonable trimming rules and to test for the sensitivity of results to these rules. In addition, I will exclude remaining outliers for which R is extremely large or extremely negative, as these observations are most likely to reflect measurement error. A small amount of trimming was done symmetrically on the basis of $hlthhat$.²⁰ In addition, since $hlthhat$ is estimated, small differences in it – and therefore in h_r^i and h_r^{i-1} – are unlikely to be significant. I therefore trim (above and substantially more below) on the basis of $hlthdif$.²¹ In addition, any remaining outliers in R (16 observations for which $R < -3$ and 16 for which $R > 10$) were excluded, again to reduce the possibility of outliers’ driving results. While I find that the exact estimate of $p(h)$ did appear somewhat sensitive to the trimming procedure used, the major conclusions of the paper appear reasonably invariant to these procedures.²²

5. Results

I first estimated the relationships for the low *Landpbig* sub-sample. The first stage estimations are shown in Table 3, Column 1. For the most part, the coefficients that were significant had the anticipated sign. Taller people, as expected, had better health, as did those with a more educated household head and no pregnant or breastfeeding household member. Being male increased health status, as did being a farmer. Age increased health but at a diminishing rate, and the positive effect of age is greater for males. In addition, having either a chronic condition oneself or an $i-1$ counterpart with such a condition reduced one’s health

²⁰ It is not strictly necessary to trim on the basis of $hlthhat$, since it is estimated using $hlthindx$, and this is defined over a compact domain. However, I opted for a conservative approach to outliers.

²¹ Observations for which $hlthdif < .15$ or > 6 were trimmed.

score significantly. Having cold temperatures (*cold*) increased health (perhaps due to the direct effect on tropical disease transmission.) Not shown are village-round dummies, most of which were individually significant at the 5 percent level, as were two ethnicity dummies. For the *hlthdif* equation, we see that cold temperatures and livestock damage to crops reduced initial differences in health, whereas the incidence of pests increased it. An index counting the number of problems in the timing and amount of rainfall (*rainprob*) increased the difference, as did having a pregnant household member (*albeit* not significantly). Having both members of the pair be female decreased initial dispersion. Individual *i*'s being a farmer increased it (and *i-1*'s being a farmer decreased it, *albeit* not significantly), as would be expected if farm work is more “productive” for the household. Finally, having the lower health member of the pair have a chronic condition increased initial dispersion significantly, as anticipated.

All second stage coefficients are reported in Table 4, column 1. The Table shows that a quadratic term in both *res* and *resdif* were selected by CV, and each of these terms is statistically significant. CV chose a fourth order polynomial in h_r^i , all terms of which were significant. In addition, having more livestock led to higher *R*, perhaps due to the labor input involved. Finally, female headship reduced relative inequality, whereas having both individuals in the pair be female increased it.²³

Figure 3 shows the estimate of $p(h)$ at mean $X\hat{\beta}$ for this sub-sample. The overfitted curve, also shown, is similar to that chosen by CV, but has a slightly lower slope.

²²That is, as long as one trims with respect to *hlthdif* and eliminates at least a few of the outliers in *R*.

²³ This result appears to somewhat contradict the first stage results, wherein initial dispersion is lowered by *bothfem*. A more complicated gender-based model (beyond the scope of this paper) may be able to explain why.

Figure 4 shows the estimate and 95 % confidence band. Clearly, the null that $\frac{\partial \hat{p}}{\partial h} = 0$ is rejected. The estimated curve is mainly upward sloping, and one cannot reject that $\frac{\partial \hat{p}}{\partial h} > 0$ everywhere (Hypothesis A.) Moreover, estimated R is for the most part less than or equal to 1. Thus, the estimate of $p(h)$ provides no evidence that productivity concerns affect the household's intertemporal allocation of health goods. It appears that households do respond, however, to equalizing preferences in allocating health goods within the household. At the same time, the coefficient on *hlthdif* is positive and statistically significant, which would not be the case under Hypothesis A. Thus, if evidence is found of a convex consumption-productivity relationship in the other sub-sample, we may conclude that productivity issues are important for this sub-sample as well, even if equality concerns dominate for the shape of $\hat{p}(h)$.

I next estimate (16) for the “high” *Landpbig* households. The first stage equations are shown in Table 3, Column 2. The effects of *cold* and *flood* on health are the same as in the low land sub-sample. Greater height, higher total assets, and greater *Landpbig* are all positively related to health. Here, having a pregnant household member is positively related to health, as is having the other ($i-1$) individual be a farmer. The presence of a chronic condition in the pair is negatively related to individual i 's initial health. Turning to initial dispersion (*hlthdif*), greater height on the part of individual i increases this dispersion, as expected. So do rainfall problems, the presence of a chronic condition, and having a pregnant household member. In addition, initial dispersion in health was greater in Round 1, when times were relatively bad. Once again sex and age matter in a complicated way through their interaction terms.

The second stage coefficients of interest are shown in Table 4. As shown, cold temperatures reduced R , (again perhaps due to the effect on disease transmission). In addition, age of the individual i is negatively related to R (with age and age squared being jointly significant), which would be consistent with older people being relatively less productive. The presence of a chronic condition for the lower health individual made equalization more unlikely, as would be expected if improving these peoples' health is relatively more costly. Table 4 also shows that movement towards greater equality was greatest on average following Round 1, when initial dispersion was greatest and health was poorest.²⁴ Therefore, it appears that households were able to improve subsequent health by allocating resources to less healthy individuals. Finally, the coefficients on the gender variables defy a simple gender (anti-female) discrimination story: Having the better health individual be male reduced R , while having both persons be female increased it.

The curve chosen by CV and the overfitted curve are shown in Figure 5, and the estimate plus 95 % confidence band are shown in Figure 6. CV chooses a downward sloping line, and once again the null hypothesis that $\frac{\partial \hat{p}}{\partial h} = 0$ can be rejected at the 5 percent level.

CV chose a quadratic in res and a cubic in $resdif$, so the endogeneity correction is once again significant.

The downward sloping shapes shown in Figures 4-5, combined with the significant portion for which $E[R] > 1$, lead to a rejection of the null of Hypothesis A for this group and provide evidence of a convexity in the consumption-productivity relationship. Moreover, the range spanned by R (from .27 to 2.5) is substantial. Thus households with moderate to high landholdings per person allocated resources in a manner consistent with strong productivity

²⁴Round-village effects were significant (and included) only for Round 1.

incentives, investing relatively more in the healthier individual than would be the case absent such concerns.

For both groups, results (Figures 4 and 6) show that households invest differentially in individuals according to their current health status, even when controlling for observable characteristics which can affect productivity and health. Equalizing incentives are in evidence particularly for low landholding households. However, given the findings for the high landholding group and the coefficient on *hlthdif* (positive and statistically significant), these households may also face a similar convexity in their consumption-productivity relationship, which is nonetheless dominated by equality concerns.

Results using BMI

I next re-estimate (16) using *bmi* to compare the results obtained for overall health to those for nutritional status. Here it was necessary to drop individuals from age 16-19 to avoid problems associated with differential growth patterns by gender.²⁵ (This also reveals a weakness of using this index as a health measure.) One could also normalize by the pooled mean *bmi* by gender before ranking individuals' health, since men in this age range tend to have a higher *bmi* than women (with a 3-round pooled mean *bmi* of 20.05 versus 19.99 for women). However, this is likely to also reflect the generally greater strength and work capacity of males. So for this exercise I interpret this higher mean weight as better "health" or work capacity, and do not normalize scores by their gender means.²⁶

²⁵When this group is included, females' mean *bmi* is higher than males'.

²⁶To obtain the sample, I trimmed the most extreme values obtained for body mass index in cases where the measured height or weight fluctuated unrealistically between rounds (9 observations out of 11 above 30). This was a general problem with these data, as heights for adults could vary from 50 to over 100 centimeters between rounds. This suggests either key coding errors or that different individuals presented themselves for measurement in different rounds. Where it was clear that a key coding error was present, I made the correction. Otherwise, I dropped the individual from the sample.

I first estimated (16) for the low *Landpbig* group. The first stage coefficients are reported in Table 3, Column 3. Here, being a farmer reduces one's body mass index, presumably because of the physical labor involved. Having had too much rain lowered *bmi* significantly but reduced initial dispersion. *Landpbig* shows a highly positive and statistically significant coefficient for initial health, but raises the initial difference in *bmi* between the individuals in the pair. The negative coefficient on *chronic* is what one would expect if these individuals received less caloric investment relative to their *i-1* counterpart.

The results for the second stage (which do not include a term in *h*) are shown in Table 4, Column 3. The coefficient on *hlthdif* is negative and significant, as would be predicted under Hypothesis A. In addition, having higher quality land increases target inequality in *bmi*. Individual *i*'s being a farmer reduces this inequality, and her being female increases it, which are not results that one would expect if one allocated net calories favoring males, whether on the basis of productivity or simply gender preference. Farmers' stores of energy are drawn down over this period, due perhaps to the effect of greater caloric expenditure, whereas females' are increased, all else equal.

The results are similar for the high landholding sub-sample. Again, I cannot reject the null that $\frac{\partial p}{\partial h} = 0$, and it appears that current nutritional status is not an important factor in intra-household allocation of net caloric investment (all else equal). The coefficients for the first stage are reported in Table 3, Column 4, and for the second in Table 4, Column 4.

These findings for *bmi* are consistent with those obtained by D&K (2000), who use the same data to test whether unpredicted working days lost due to unexpected illness affect subsequent changes in *bmi*. For most groups, they do not reject the null of no effect (*i.e.*,

In addition, I dropped observations for which *hlthdif* was less than .1 (numbering 17) or greater than 8 (14).

efficiency in their model of perfect risk sharing). According to the results obtained here, households do not on average allocate net calories on the basis of initial *bmi* either (except with regard to initial dispersion).²⁷ These results do contrast somewhat with those of Pitt, *et al.* (1990), however. Examining the intra-household allocation of net calories in Bangladesh, they find that due to productivity complementarities with effort, people with intrinsically higher weight-for-height are allocated more calories. But since these people also expend more calories, they are in effect taxed to improve the nutritional status of others. Thus equalization is preferred on average. In rural Ethiopia when nutritional status is measured as *bmi*, however, initial *bmi* does not affect the subsequent allocation of calories. While farmers (and possibly males) appear to be “taxed” in terms of net calories, on average equalization is not preferred. Not only is $\hat{p}(h)$ not increasing in *bmi*, but average *R* is not less than 1 (at 1.03).

6. Conclusion

The results of this paper lend support to a model of buffer health stocking with an important equality-productivity tradeoff in the allocation of health investment, particularly for households with relatively high landholdings per working age member. Under efficiency, these households must weigh future productivity in their overall health investments. The very desire to equalize health itself further reduces household income in the short run; likewise, the desire to increase productivity will impede an ill-health person’s recovery. Thus, the presence of a convexity in the consumption-productivity relationship means that

²⁷The results for *hlthhat* may lead one to question one of D&K’s key identifying assumptions, however. As they point out, under optimality households must take account of productivity issues in their allocation of nutrition (or health status.) Here, even household (or village-) level shocks to health will affect individual productivity – and therefore the “user cost” of health – differently within the

adverse income and health shocks can reduce productivity for some time, possibly producing a multiplicity of equilibria.

These results are based on a relatively short time frame, during which times were initially difficult, *albeit* not disastrous for households possessing land, and then improved. In times of greater distress or famine, the dynamics could be qualitatively different. Thus, a longer panel would be needed to examine the extent to which the effects of poor health (or income) shocks endure, particularly following a famine. As suggested by earlier work, such downturns may have much more persistent effects on the long term health and work capacity of children.

Finally, these results emphasize the importance of moving beyond pure body-mass - based measures of health status if important dynamics are to be identified. Conducting the test proposed here using such a measure would obscure important dynamics in the relationship between consumption and productivity. These were uncovered in these data only when a broader measure of health status was used.

household due to the non-linearities in $y(\tilde{g})$. This may not be the case with nutrition, however. D&K must assume not for their test to be valid.

Table 1: Task Measures by Sex, Working Age (Pooled)

	Male		Female	
	Mean	St. Dev.	Mean	Std. Dev.
<i>Stand Up</i>	3.957	.2739	3.924	.3555
<i>Sweep</i>	3.955	.3017	3.927	.3769
<i>Walk</i>	3.947	.3412	3.867	.5137
<i>Carry</i>	3.889	.4901	3.753	.7057
<i>Hoe</i>	3.886	.5061	3.506	.9874

Table 2: Distribution of *Hlthindx*, *Hlthhat*

<i>Hlthindx</i>	Full Sample (1)	(%)	<i>Hlthhat</i>	Full (3)	(%)	Used	(%)
1	30	(.29)	0-1	13	(.17)	0	(0)
2	18	(.18)	1-2	60	(.76)	0	(0)
3	30	(.30)	2-3	159	(2.02)	0	(0)
4	34	(.33)	3-4	281	(3.58)	15	(1.04)
5	25	(.25)	4-5	418	(5.23)	40	(2.77)
6	44	(.43)	5-6	428	(5.45)	63	(4.36)
7	68	(.67)	6-7	43	(.55)	2	(.14)
8	68	(.67)	7-8	1242	(15.81)	204	(14.13)
9	62	(.61)	8-9	2378	(30.27)	548	(37.95)
10	123	(1.21)	9-10	1886	(24.01)	386	(26.73)
11	159	(1.56)	10-11	803	(10.26)	158	(10.94)
12	132	(1.29)	11-12	125	(1.59)	28	(1.94)
13	366	(3.59)	12-13	4	(.05)	0	(0)
14	221	(2.17)	13-14	4	(.05)	0	(0)
15	331	(3.25)	14-15	6	(.08)	0	(0)
16	8475	(83.20)	15-16	1	(.01)	0	(0)
			16-17	1	(.01)	0	(0)
	10186			7855		1444	

Table 3: First Step Estimates (Robust Standard Errors in Parentheses)

Measure:	<i>Hlthhat</i> Low land		<i>Hlthhat</i> High land		<i>Bmi</i> Low land		<i>Bmi</i> High Land	
	(1)	(1)	(2)	(2)	(3)	(3)	(4)	(4)
	h_r^i	$h_r^i - h_r^{i-1}$	h_r^i	$h_r^i - h_r^{i-1}$	h_r^i	$h_r^i - h_r^{i-1}$	h_r^i	$h_r^i - h_r^{i-1}$
<i>Totasset</i>	--	--	.00001** (.00000)	.0001* (.0001)	--	-.00007** (.00002)	--	.000024 (.000016)
<i>SalVal</i>	--	--	--	--	--	-.00001 (6.61e ⁻⁶)	--	--
<i>Height</i>	2.75** (.347)	.608 (.508)	3.56** (.360)	1.26** (.638)	--	--	--	--
<i>Too much</i>	--	--	--	--	-.705** (.223)	-.368** (.160)	--	.318 (.206)
<i>Landpbig</i>	--	--	.0038** (.0013)	--	34.43** (16.05)	23.39* (13.15)	--	--
<i>Rainprob</i>	--	.099** (.031)	--	.146** (.066)	--	--	--	.276 (.224)
<i>Cold</i>	.621** (.112)	--	.659** (.216)	--	--	--	1.37** (.584)	--
<i>Flood</i>	-.279** (.119)	-.344* (.174)	-.312** (.108)	--	.451 (.278)	--	-1.03** (.354)	--
<i>Pests</i>	--	.216* (.126)	--	--	--	--	--	--
<i>PcLem</i>	--	--	--	-.161 (.123)	--	--	--	-.640** (.219)
<i>Pregghm</i>	-.121* (.064)	.143 (.089)	.149** (.061)	.188* (.111)	--	--	--	--
<i>RI</i>	--	--	--	.259** (.096)	-2.95** (.604)	--	-4.16** (.468)	--
<i>Age</i>	.040** (.013)	.020** (.007)	-.049** (.004)	.025** (.007)	-.0097 (.0078)	--	.126** (.046)	.109** (.039)
<i>Age²</i>	-.0016** (.000)	.0009** (.0003)	--	--	-.0014** (.0005)	--	-.0016** (.0006)	-.001** (.0005)
<i>Ageⁱ⁻¹</i>	--	-.028 (.019)	--	-.027 (.020)	.100** (.043)	--	.073* (.042)	.051 (.035)
<i>Ageⁱ⁻¹²</i>	--	.001** (.0002)	--	.0010** (.0003)	-.0015** (.0005)	--	-.001** (.0005)	-.0007 (.0004)

<i>Male</i>	.365** (.164)	--	--	--	--	--	--	--
<i>Bothfem</i>	--	-1.25** (.210)	.371** (.129)	-.436* (.242)	--	--	--	--
<i>Male*age</i>	.020** (.006)	-.025** (.012)	.068** (.007)	.004 (.014)	--	-.016** (.004)	-.055** (.019)	-.007* (.004)
<i>Male*ageⁱ⁻¹</i>	--	-.012 (.011)	-.011* (.006)	-.031** (.005)	--	--	--	.0094 (.0065)
<i>Male*age²</i>	--	-.0004 (.0002)	-.0009** (.0001)	-.0007** (.0002)	--	--	.0007* (.0004)	--
<i>Male*ageⁱ⁻¹²</i>	--	-.0004** (.0002)	.0002** (.0001)	-.031** (.005)	--	--	--	--
<i>Farmer</i>	.204** (.078)	.217* (.114)	--	.360** (.147)	-.725** (.218)	--	--	--
<i>Farmerⁱ⁻¹</i>	--	-.254 (.161)	.269** (.102)	.413** (.182)	--	-.371** (.146)	-.617** (.210)	-.914** (.281)
<i>Head</i>	--	.378** (.136)	--	--	.806** (.269)	.946** (.206)	--	--
<i>Headⁱ⁻¹</i>	--	.326** (.136)	--	--	.788** (.255)	.772** (.182)	--	--
<i>Totsmall</i>	.103** (.016)	--	--	--	.075* (.044)	--	.173** (.045)	--
<i>Chronic</i>	-.185** (.086)	--	-.187** (.079)	.251* (.141)	-.395* (.232)	-.364** (.178)	.348 (.231)	--
<i>Chronicⁱ⁻¹</i>	-.184** (.074)	.444** (.098)	-.183** (.066)	.647** (.114)	--	--	--	--
<i>Trampled</i>	--	-.321** (.153)	--	--	--	--	--	--
R ²	.7526	.2830	.8225	.3138	.3348	.1343	.3203	.1942
N	843	843	601	601	669	669	468	468

Notes: Round interacted with village for a round included if jointly significant. Only individually significant ethnicity dummies included. *Significant at 10 percent. **Significant at 5 percent.

**Table 4: Best Semi-Parametric Instrumental Variables Results for Equation (16)
(Adjusted, White-Corrected Standard Errors in Parentheses)**

<i>Measure:</i>	<i>Hlthhat Low Land</i> (1)	<i>Hlthhat High Land</i> (2)	<i>Bmi Low Land</i> (3)	<i>Bmi High Land</i> (4)
$h_r^i / h_r^{i^2} / h_r^{i^3} / h_r^{i^4}$	-7.74(4.74)/ 1.70*(.971)/ -.148*(.085)/ .0045*(.0027)	-.272**(.0996) -- -- --	--	--
$h_r^i - h_r^{i-1}$.1038*(.058)	.201** (.076)	-.285* (.155)	-.540** (.134)
<i>Resdif</i>	-.608**/.105**	-.618**/.049*	.006/.044**	.228*/.177**/-.019**
<i>Resⁱ</i>	-.301**/.128**	.311 / -.043 /-. .014	--	--
<i>SalVal</i>	1.56xe ⁻⁷ ** (6.99 xe ⁻⁸)	--	--	.00001** (4.98e ⁻⁶)
<i>Totasset</i>	--	--	--	.0003** (9.23e ⁻⁶)
<i>PcLem</i>	--	--	.358** (.176)	--
<i>Cold</i>	--	-.364** (.132)	--	--
<i>Age/agesq</i>	--	-.0237(.027)/- .0006(.0004)	--	-.015** (.007)
<i>Round 1</i>	--	-1.32** (.236)	--	--
<i>Male</i>	--	-.769** (.306)	-.437* (.248)	--
<i>Both Female</i>	.486** (.165)	.292 (.199)	--	.457 (.354)
<i>Farmer</i>	--	--	-.470** (.225)	--
<i>Farmerⁱ⁻¹</i>	--	--	--	-.445** (.216)
<i>Femhead</i>	-.813** (.309)	-.319 (.229)	--	-.445** (.216)
<i>Totsmal</i>	--	-.0388 (.0245)	--	--
<i>Chronic</i>	--	--	--	.432** (.201)
<i>Chronicⁱ⁻¹</i>	--	-.365**	.282	--

		(.146)	(.181)	
<i>Male*age</i>	-0.002	.052** (.010)	.0087 (.0058)	--
<i>Male*ageⁱ⁻¹</i>	--	--	-.006 (.004)	.040** (.014)
<i>Male*ageⁱ⁻¹²</i>	--	--	--	-.00047** (.0002)
<i>Illdth</i>	--	--	.161 (107)	--
N	843	610	669	468

See Notes, Table 3.

Figure 1

Hypothesis A

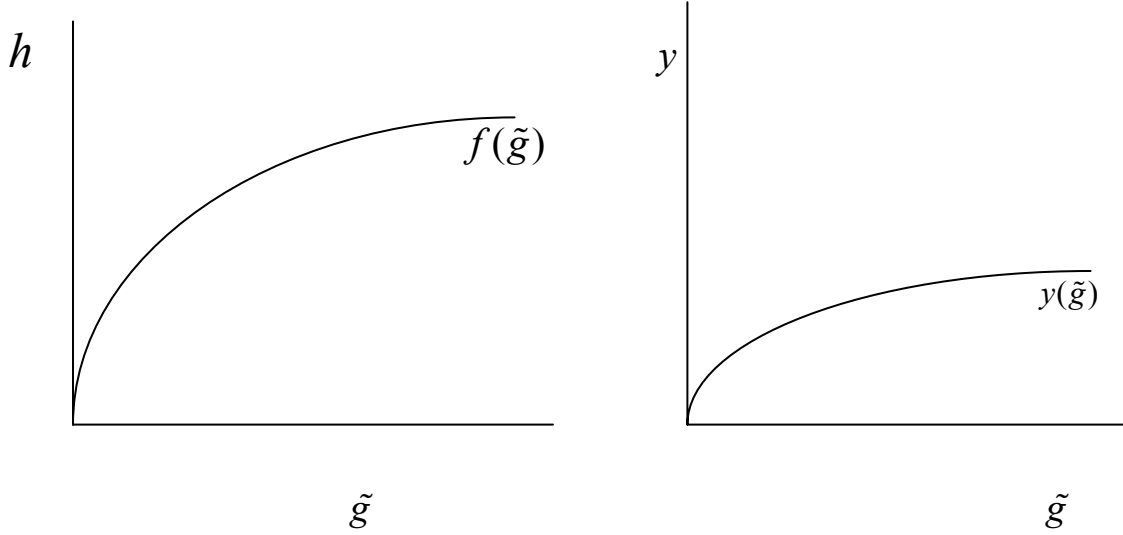


Figure 2

Hypothesis B

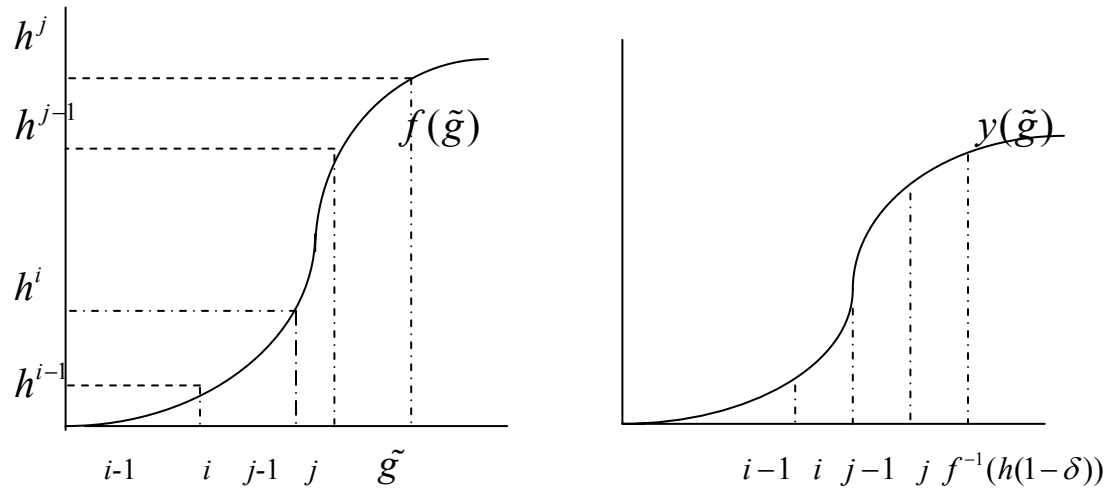


Figure 3

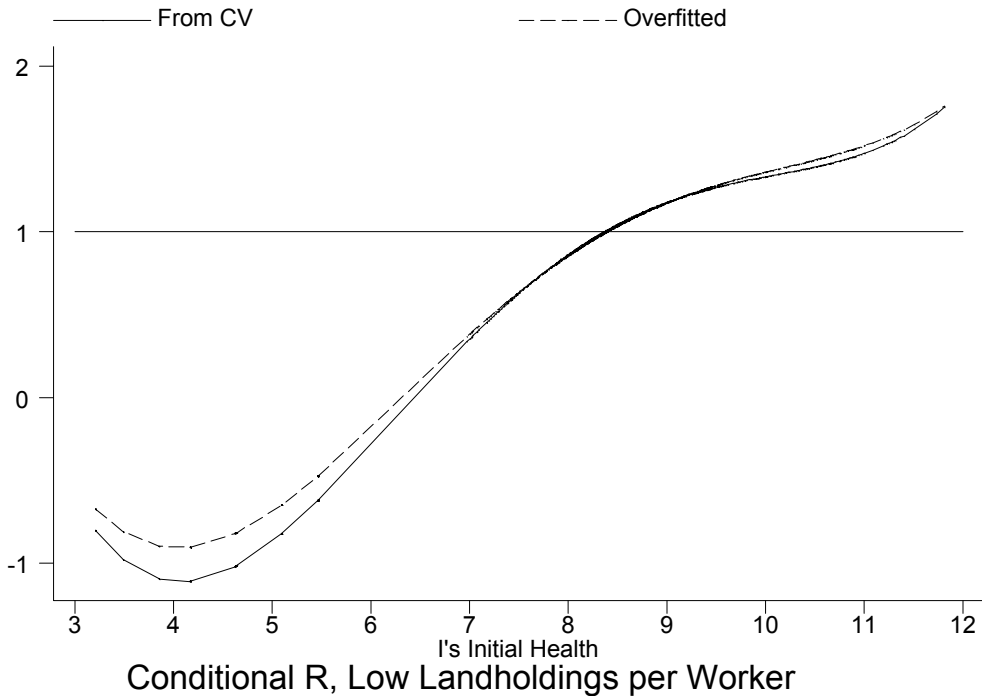


Figure 4

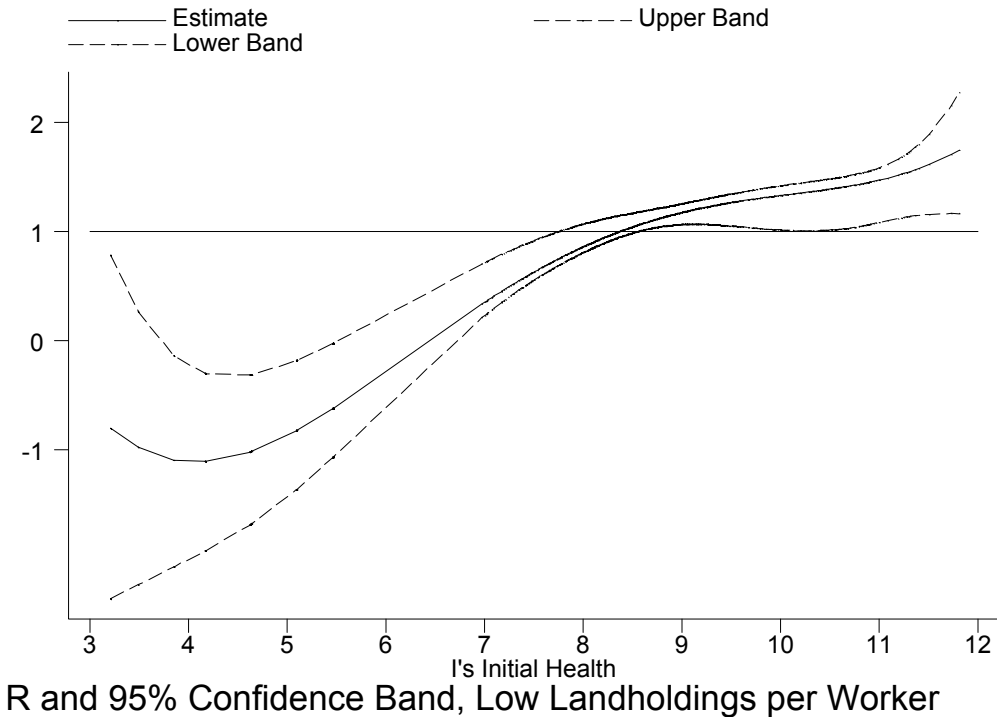


Figure 5

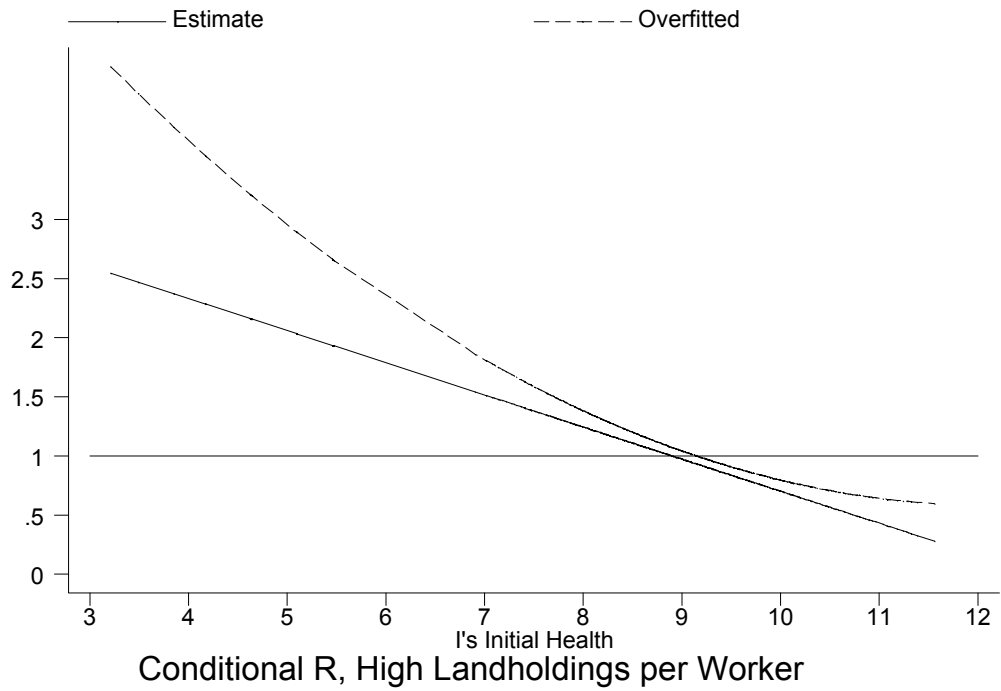
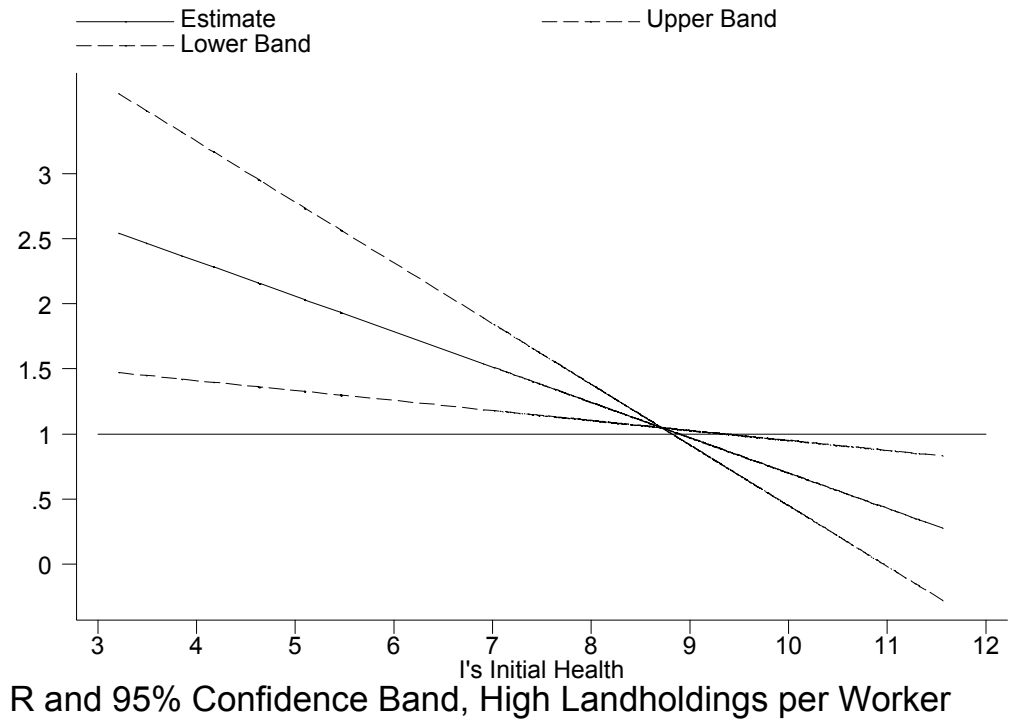


Figure 6



APPENDIX 1

Proof of Proposition 1.

First, note that according to standard results in Deaton (1991), given our assumptions on the utility function (6), we know that:

$$(a) \frac{\partial V}{\partial h} < 0 \text{ and } \frac{\partial^2 V}{\partial h^2} > 0 .$$

We know also from (11) that $\frac{\partial V}{\partial h} > \frac{\partial u}{\partial h}$ for all h . Thus, from (10 (a)), we have that:

$$(b) \frac{\partial V_t}{\partial h_t^{i-1}} - \frac{\partial u}{\partial h_t^{i-1}} > \frac{\partial V_t}{\partial h_t^i} - \frac{\partial u}{\partial h_t^i}, \text{ which implies that}$$

$$(c) E_t \left[\frac{\partial V_{t+1}}{\partial X_{t+1}} \frac{\partial y_{t+1}}{\partial h_t^{i-1}} + (1-\delta) \frac{\partial V_{t+1}}{\partial h_t^{i-1}} \right] > E_t \left[\frac{\partial V_{t+1}}{\partial X_{t+1}} \frac{\partial y_{t+1}}{\partial h_t^i} + (1-\delta) \frac{\partial V_{t+1}}{\partial h_t^i} \right].$$

Since $\frac{\partial y_{t+1}}{\partial h_t^{i-1}} > \frac{\partial y_{t+1}}{\partial h_t^i}$ and $E_t \frac{\partial V_{t+1}}{\partial h_{t+1}}$ is monotonically decreasing in $E_t h_{t+1}$, there are two

possibilities. First, it could be the case that $E_t h_{t+1}^{i-1} > E_t h_{t+1}^i$. It must follow from this that

$g_t^i < g_t^{i-1}$. Alternatively, if $E_t h_{t+1}^{i-1} < E_t h_{t+1}^i$, and if $\frac{\partial f}{\partial \tilde{g}_t^i} < \frac{\partial f}{\partial \tilde{g}_t^{i-1}}$ then from (9) it must be the

case that $g_t^i < g_t^{i-1}$. If $\frac{\partial f}{\partial \tilde{g}_t^i} \geq \frac{\partial f}{\partial \tilde{g}_t^{i-1}}$, then it follows directly that $g_t^i < g_t^{i-1}$. It follows from

$g_t^i < g_t^{i-1}$ and the concavity of $f(\tilde{g})$ that $E_t \left[\frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}} \right] < 1$.

Proof of Proposition 2:

Suppose that there are some individuals i, j such that $h_t^j > h_t^i$ and their nearest neighbors in the health distribution, $i-1, j-1$, are such that $h_t^j - h_t^{j-1} = h_t^i - h_t^{i-1}$. Suppose that the

household wishes to set $E_t \left[\frac{h_{t+1}^j - h_{t+1}^{j-1}}{h_t^j - h_t^{j-1}} \right] < E_t \left[\frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}} \right]$. This would imply that

$g^{j-1} - g^j > g^{i-1} - g^i$. Given proposition 1, we know that $g^j < g^{j-1} < g^i < g^{i-1}$. These conditions combined plus the concavity of the utility function imply

$u'(g^j) - u'(g^{j-1}) > u'(g^i) - u'(g^{i-1})$. This can be rewritten as

$u'(g^{j-1}) - u'(g^j) < u'(g^{i-1}) - u'(g^i)$. From Equation (9) this implies that

$\frac{\partial f}{\partial \tilde{g}^{j-1}} EV_{h_{t+1}^{j-1}} - \frac{\partial f}{\partial \tilde{g}^j} EV_{h_{t+1}^j} > \frac{\partial f}{\partial \tilde{g}^{i-1}} EV_{h_{t+1}^{i-1}} - \frac{\partial f}{\partial \tilde{g}^i} EV_{h_{t+1}^i}$. As long as the household does not

wish to change the order of the health rankings, this violates the joint concavity of $f(\tilde{g})$ and $V(., h)$. Thus, a contradiction arises, and it must be the case that the household wishes to set

$$E_t \left[\frac{h_{t+1}^j - h_{t+1}^{j-1}}{h_t^j - h_t^{j-1}} \right] > E_t \left[\frac{h_{t+1}^i - h_{t+1}^{i-1}}{h_t^i - h_t^{i-1}} \right].$$

APPENDIX 2

The general form of the ordered model is well known. A latent continuous variable for health, h^* is assumed to underlie the discrete ordered measure, and the goal here is to estimate the threshold points which determine the latent variable's assignment to the discrete values. For example, for which $hlthindx=j$ where $\mu_{j-1} < h^* \leq \mu_j$. The estimated threshold points are taken as estimates of the "true" distance between (upper) values for $hlthindx$. I estimate the following model by ordered probit:

$$hlthindx_t^i = \alpha + \pi W_t^i + u_t^i,$$

where u and W are assumed to be independent. I use the pooled (3 rounds of) data for all individuals from 16 to 64 for whom the data were complete. The W variables were height (a long run measure of health, age, age squared, sex, sex interacted with age, and this variable squared, whether the individual was mainly engaged in farm work, the household's landholdings per working age person, the fraction of landholdings of highest quality, the number of individuals having died of illness in the previous 5 years, the total number of teenage and adult persons, the total number of children, the time it takes to collect water, the total value of household assets as of Round 1, the education of the household head, dummy variables for whether there was too little rain in the last main harvest season, whether there was too much, whether the rains stopped late, and an index of other rainfall problems. In addition, I included a set of dummies for other harvest related events, such as flooding, cold temperatures, and insect infestation, and Round-village dummies for all village-round combinations. There were 9298 observations, and the estimation had a Likelihood Ratio of 2106. The measures as constructed according to equation (12) (and transformed linearly by multiplying by 2 and adding 4.148) produced values that preserved the original rankings,

except for where $hlthindx=1$ and $hlthindx=16$. The coefficients and threshold values are as shown in Table A.

Table A: Results of Ordered Probit Estimation of *Hlthindx*

	Coefficients	(s.e.)	Cut Points	(s.e.)
<i>Toolittle</i>	-.252**	(.078)	-1.65	(.472)
<i>Toomuch</i>	-.180**	(.073)	-1.45	(.470)
<i>Stoplate</i>	-.116*	(.060)	-1.22	(.468)
<i>Rainprob</i>	.106**	(.046)	-1.04	(.467)
<i>Cold Temp</i>	.253*	(.084)	-.933	(.467)
<i>Wind/Storm</i>	.072	(.064)	-.827	(.466)
<i>Flood</i>	-.235**	(.073)	-.662	(.466)
<i>Pests</i>	-.069	(.059)	-.522	(.466)
<i>Insects</i>	.029	(.070)	-.416	(.466)
<i>Trampling</i>	.105	(.068)	-.247	(.466)
<i>Landpbig</i>	.0022	(.0018)	-.055	(.466)
<i>PcLem</i>	-.0002	(.0003)	.076	(.466)
<i>Totasset</i>	.00005**	(.00002)	.395	(.466)
<i>Illdth</i>	-.007	(.025)	.551	(.466)
<i>Pregghm</i>	.024	(.038)	.761	(.466)
<i>Eduhead</i>	.014	(.010)		
<i>Farmer</i>	.089	(.066)		
<i>Age</i>	-.002	(.009)		
<i>Age²</i>	-.0004**	(.0001)		
<i>Male</i>	-.011	(.258)		
<i>Male*age</i>	.029*	(.015)		
<i>Male*age²</i>	-.0004*	(.0002)		
<i>Totbig</i>	-.002	(.008)		
<i>Totsmall</i>	.037**	(.010)		
<i>Height</i>	1.87**	(.199)		
N	9298			

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