Soil Investments on Rented versus Owned Plots: Evidence from a Matched Tenant-Landlord Sample in Malawi

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ABSTRACT We use a unique data set on matched tenant-landlord pairs in Malawi to compare decisions on smallholder plots that were rented versus those that were owner-operated. Controlling for household and rental-pair fixed effects, we found that some input use (e.g., hybrid maize seed) and soil fertility investments (e.g., manure, compost, minimum tillage) were higher on tenants’ owner-operated plots than on their rented-in plots. Tenants were also less likely to use compost than their landlords. Landlords were less likely to rent out plots with fruit trees. Our results suggest that the expansion of farmland rental markets may exacerbate soil fertility maintenance concerns. (JEL D63, O12)

1. Introduction

Recent growth of farmland rental markets in sub-Saharan Africa (SSA) can be understood as part of an ongoing structural transformation process in the region. In the foundational discussions of this process, Johnston and Kilby (1975) explain how higher returns to labor in nonfarm sectors induce an exodus of labor out of agriculture, with relatively more efficient farmers remaining in agriculture, expanding their capital and land endowments. This self-selection contributes to sectoral productivity gains, which are also driven by investments incentivized by increasing food demand and improving input market conditions. During this process, land transfers through rentals and sales should facilitate the reallocation of land resources to these more efficient farmers. For example, functioning land rental markets should allow tenants to expand the cultivated area and bring more capital into the sector while potentially providing landlords with compensation for their land assets while they engage in pursuits outside of agriculture.

Although most land cultivated by smallholders in SSA is managed in customary tenure systems where operators lack formal titles, recent evidence from the region suggests growth in land rental markets has been pronounced (Holden, Otsuka, and Place 2009; Chamberlin and Ricker-Gilbert 2016). Chamberlin and Ricker-Gilbert (2016) found that the percentage of households renting in land rose from 7.5% in 2002–2003 to 15.4% in 2008–2009 in Malawi and from 0.9% in 2001–2002 to 3% in 2012–2013 in Zambia. This suggests that growth in rental market participation has been taking place even in environments where tenure rights are still largely customary and tenure security is sometimes ambiguous.

At the same time, much of the region is facing a soil fertility crisis, with widespread soil nutrient losses occurring from year to year due to soil erosion and nutrient mining (Sanchez 2002). The concurrent trends of increasing land rental participation and decreasing soil
fertility are worth scrutinizing because there is some evidence to suggest that rented plots receive lower soil fertility–enhancing investments than owner-cultivated plots. For example, Gavian and Fafchamps (1996), Jacoby and Mansuri (2008), Lovo (2016), and Murakoka, Jin, and Jayne (2018) find that tenants are typically less likely to make soil-enhancing investments (such as animal manure application) on plots that they rent compared with plots they own and cultivate. This is a rational decision by tenants, given that rental contracts tend to be short-term, while the benefits of applying animal manure may take multiple years to be fully realized.

The present study adds to the literature on soil fertility and land renting using a unique plot-level data set from Malawi, collected in 2016, that matched tenants with landlords and collected soil samples on rented and owner-operated plots. Specifically, for each tenant we observed their rented-in plots (tenant’s rented-in plots) and all their owner-operated plots (tenant’s owner-operated plots). One unique aspect of this data set is that we also surveyed tenant’s respective landlords, and thus we also observed the plots that the landlord did not rent out (landlord’s owner-operated plots). The other novel feature of these data are that we took objective soil fertility measures including nitrogen (N), phosphorous (P), potassium (K), pH, organic matter, silt, clay, and sand on tenants’ largest owner-operated plot, their largest rented-in plot, along with the largest plot that was owner-operated by the landlord who owned the largest rented-in plot. This allowed us to expand our estimates of soil quality beyond the self-reported indicators used in most studies (usually if the soil on the plot was of good, fair, or poor quality).

Using this unique data set, we answered the following research questions related to land rental markets in SSA, which have been understudied to date. First, did input use (e.g., labor for weeding, applying herbicides, applying inorganic fertilizer, and planting hybrid maize) and soil fertility investments (e.g., maize and legume intercropping, applying manure, applying green compost, using minimum tillage) differ on average among (1) a tenant’s rented-in plots, (2) a tenant’s owner-operated plots, and (3) the owner-operated plots of the tenant’s landlord? As mentioned, much of the previous literature indicated that tenants were less likely to make longer-term soil fertility investments on their rented-in versus owner-operated plots. However, landlord information was missing from these studies, limiting our insights into managerial changes that occur when land was transferred from landlords to tenants. The present study seeks to understand how a tenant’s input use and soil fertility investments compared to their landlord’s investments on their own plot. To our knowledge, this has not been empirically tested before using a matched tenant-landlord data set.

Second, we ask: did landlords rent out plots that they believe to be of lower or higher soil fertility ex ante than those they cultivated themselves, and how was this affected by the landlord’s perception of tenure security on the plot? We might expect a landlord to have rented out land of lower soil fertility, ex ante, if he or she felt that such selection would not adversely affect rental receipts, or if he or she felt that renting out customary land may have increased the risk of having that land reallocated to someone else by a village leader. However, tenants generally have been found to be wealthier and better educated than landlords in our study area (Ricker-Gilbert et al. 2019). This suggests potential power
imbalances, possibly giving tenants an advantage over landlords when selecting which plots to rent in. We therefore ask: were land rental transfers more likely to be from landlords’ higher or lower quality plots?

By combining a matched tenant-landlord data set with objective and subjective measures of soil fertility, the present article makes an important contribution to the land rental market and land tenure literature in SSA. Several earlier studies used matched tenant-landlord data sets, but with different objectives. For example, Deininger, Ali, and Alemu (2013) used a matched sample to estimate the relative differences in Marshallian efficiency between share-cropped plots, plots rented at a fixed rate, and owner-operated plots in Ethiopia. Bellemare (2012) used a matched sample from Madagascar to estimate how a landlord’s perception of his or her tenure security affects the choice of contract offered to tenants. Ghebru and Holden (2014) used a matched landlord-tenant sample from Ethiopia to assess bargaining power, efficiency, and distributional implications in tenant-landlord relationships. Ricker-Gilbert et al. (2019) use the same matched data set in Malawi as in this study to estimate efficiency and equity trade-offs between tenants and landlords when land rental arrangements are made. Ours is the first to use a matched tenant-landlord sample to estimate models related to land renting, soil fertility, and other investments.

Other studies have used objective measures of soil fertility to estimate how these measures relate to farmers’ subjective self-assessment of soil fertility. Berazneva et al. (2018) with data from Kenya and Tanzania and Gourlay et al. (2017) with data from Ethiopia and Uganda find that the link between objective (laboratory test-based) assessments and subjective (farmer reported) assessments of soil fertility is weak and that farmers base their subjective assessment on a plot’s crop yield. To our knowledge, the present article is the first to use objective measures of soil fertility to understand how landlords use (or do not use) such information to make land use decisions.

To answer our research questions, we estimated two econometric models. In the first model, we regressed input use and soil fertility investments on land rental status for tenant’s and landlord’s plots, along with household and plot-level controls using linear models. The identification strategy for these models relied on the matched tenant-landlord sample, allowing us to use rental-pair fixed effects (FEs). Second, we estimated a linear probability model (LPM) for landlord households only and regress a binary variable equaling one if the landlord rented out a specific plot on the following factors: (1) objective soil fertility measures on the plot, (2) subjective soil fertility measures on the plot, (3) landlord perceptions of tenure security on the plot, and (4) investments made on the plot. In this model, our identification strategy was based on the fact that we have soil samples for multiple plots in the landlord household (rented out and owner cultivated). This allows us to use a within-household FE to control for unobserved factors associated with the landlord that may affect her rental decision. We also had a rich set of plot-level information in our data set that should control for the vast majority of remaining unobserved factors that might bias the coefficient estimates in our models. That being said, as with any observational study, our causal identification claims are made cautiously. Nonetheless, we believe that the analysis uncovers important relationships that are useful for smallholder agricultural policy in SSA.

Briefly, our results indicate that tenants in our sample were more likely to use inputs like inorganic fertilizer on plots they rented in and plots they owner-cultivated themselves, compared with their landlords. However, tenants were less likely to make soil fertility–enhancing investments such as green compost on their rented-in plots compared with what their landlords did on their owner-operated plots. Tenants were significantly less likely to invest in applying animal manure, green compost, or minimum tillage on their rented-in plots compared with what their landlords did on their owner-operated plots. Tenants were significantly less likely to invest in applying animal manure, green compost, or minimum tillage on their rented-in plots compared with what their landlords did on their owner-operated plots on average. Furthermore, it seems that the major factors affecting whether a landlord rented out a plot or cultivated it herself was whether there were observable assets such as fruit trees on the plot. Landlords were 49%–51% less likely to rent out a plot with a fruit tree on it, probably due to the income and nutrition that the fruit provided them.
2. Land Rental Markets in Malawi

Our data and analysis are set in Malawi. The country officially recognizes three types of land tenure regimes: (1) the public tenure system, which includes public lands, national parks, and so on; (2) the private tenure system, which includes freehold, and leasehold land, where owners have titles and land can be bought, sold, and rented; and (3) the customary tenure system. The customary tenure system is by far the largest in Malawi in terms of landholding and number of people engaged (Lunduka, Holden, and Øygard 2009). Households who cultivate in the customary system have no formal title but have user rights granted by local chiefs. These land rights may be passed down from parents to children but are ultimately controlled by the chief. Land renting is not explicitly allowed in the customary system, but land rental participation has been growing in Malawi, as mentioned earlier, under the de facto approval of local leaders.

Recent evidence suggests that land rental contracts in Malawi are almost entirely up-front cash-rent and short-term in nature, lasting one to two years on average (Ricker-Gilbert et al. 2019). In addition, tenants in Malawi have been found to be significantly wealthier than landlords along all dimensions of wealth other than land. This includes assets, savings, education, and available labor (Chamberlin and Ricker-Gilbert 2016; Ricker-Gilbert et al. 2019). Chamberlin and Ricker-Gilbert (2016) and Ricker-Gilbert et al. (2019) found evidence to support the notion that landlords rented out land because of the need for immediate cash or because they lacked the necessary labor, rather than to earn money that could facilitate engagement in other productive nonfarm activities.

3. Land Tenure Security and Soil Fertility

The literature clearly documents that land tenure insecurity is prevalent in many parts of Africa and in Malawi in particular. For example, in a survey of evidence across southern Africa, Mutangadura (2007) found that major causes of tenure insecurity included loss of land rights for minority groups, unclear or overlapping land rights, overcrowding, and land alienation into leasehold, among other reasons. In particular, the author found that the transfer of land from customary to leasehold for investors during the 1980s and 1990s and high and increasing population to land ratios were two of the major drivers of tenure insecurity in Malawi. A recent study in Malawi found that tenure insecurity is widespread, with 22% of land users being concerned about losing their land (Deininger, Xia, and Holden 2019). Peters and Kambewa (2007) found that impending land reform legislation that would formalize tenure rights in the customary sector in Malawi led to increased tenure insecurity and competition for land among smallholders.

Other studies from Malawi have found that people who lived in the home area of their spouses (e.g., a man residing in his wife’s village) felt more tenure insecure (Lunduka, Holden and Øygard 2009; Matchaya 2009). Both Lunduka, Holden, and Øygard (2009) and Matchaya (2009) pointed out that particularly when a man resided in the wife’s village and had no claim to the land other than through his wife, he was less likely to make longer-term investments in soil fertility. This was because he had no claim to the land if the marriage ended or his wife died.

Previous literature also supported the view that farmers have greater incentives to maintain soil fertility on owner-operated plots than on rented-in plots. This has been attributed to lack of commitment from tenants because of the short-term nature of most land rental contracts (Jacoby and Mansuri 2008). Jacoby and Mansuri (2008) found that in Pakistan, tenant farmers were less likely to invest in a noncontractible soil fertility investment (e.g., applying animal manure) than on owner-operated plots. This was even the case on plots under fixed rent, where the tenant maintained all output, compared with share-cropped plots (where the tenant shared output with the landlord). In the African context, Gavian and Fafchamps (1996; for Niger), Lovo (2016; for Malawi) and Muraoka, Jin, and Jayne (2018; for Kenya) all found that rented-in plots were less likely to receive soil conservation or soil
fertility–enhancing investments compared with owner-cultivated plots.

The studies cited here provided important context that motivated the present work. We build on this literature by using a matched sample of tenants and landlords to answer questions about input use and investments made on a tenant’s rented-in plot, their owner-operated plot, and a landlord’s owner-operated plot. We also seek to understand how objective and subjective measures of soil fertility and other investments influence a landlord’s decision to rent out a particular plot versus cultivate it for themself.

4. Data

Data in this study came from four districts in Malawi collected between April and June 2016, which immediately followed the 2015–2016 season. These districts were Lilongwe, Salima, and Nkhotakota in the central region, and Zomba in the southern region. Nkhotakota and Salima were selected to represent rural areas, and Lilongwe and Zomba were selected to represent peri-urban areas. These specific districts were selected because of their high incidence of land rental market activity according to the 2009–2010 nationally representative LSMS-IHS3 data. The total target sample size was 600, representing 150 farm households per sampled district. Since the IHS3 was only representative at the district level, after using those data to select districts, we contacted the District Agricultural Development Officer (DADO) as the entry point for the survey team. We used the DADO’s local knowledge to identify Extension Planning Areas (EPAs) with high rental market activities in each district. In these EPAs, we randomly selected villages for our sample, choosing one village per EPA.

Once a village was selected, the field supervisors along with the local extension officer undertook a targeted household listing exercise. Smallholder farm households participating in land renting were identified through a focus group discussion (FGD) with the village headman (local leader, also known as ‘chief’), lead farmers, and members of village development committees and the vulnerability assessment committee. These individuals were considered to be knowledgeable about the history of the village and related land issues. On average, the FGD comprised about 10 people, of which half were women. Community-level issues regarding landownership, land availability and use, drivers of land renting, and prevailing farm gate prices of cash crops were discussed. The use of key informants to help identify respondents to answer questions about sensitive land-related issues has been used in many previous studies, including Bardhan and Mookherjee (2010); Macours, de Janvry, and Sadoulet (2010), Vranken et al. (2011), and Macours (2014).

At the end of each FGD, we randomly sampled individual farming households from the village list of all households to serve as a sampling frame for our survey.4 The village leaders helped us identify whether the randomly selected households were tenants, landlords, or autarkic (neither renting in or renting out land). Each sampled landlord was matched to his or her tenant as pairs for the household interviews. Thus, if a tenant household was sampled, its corresponding landlord was automatically sampled for the interview and vice versa.5 Households were selected with replacement until a sample size of 10 matched pairs was reached (i.e., 20 households). Furthermore, 10 autarkic households were randomly selected from the list as control households. Thus, a total of 30 households were sampled per village.

After cleaning the household data, our data set consisted of 948 unique plots from 169 tenant and 169 landlords who formed rental pairs. Of these 948 plots, 347 were tenant’s rented-in plots, 266 were tenant’s owner-operated plots, and 335 were landlord’s owner-operated plots. Appendix Table A1

4 Village lists in Malawi were regarded as being accurate because they were used to determine how many input subsidy vouchers were given to a particular community and then to households in the community. Therefore, households had incentives to make sure they were included on the list.

5 Tenants and landlords formed unique pairs in this analysis. If at tenant (landlord) had multiple landlords (tenants) then only the landlord (tenant) who owned (operated) the largest rented plot was found for interview. This was done for logistical purposes to keep the time and duration of the survey manageable.
presents the frequency on the number of plots controlled by tenants and landlords in the data.

**Soil Data**

We took soil samples and GPS estimates of the area of surveyed plots according to the following criteria: the largest owner-operated plot cultivating maize or other annual crops for tenants and landlords. We took soil samples and GPS estimates of the area for the largest rented-in plot for each tenant and matched it to the landlord who had rented it out during that season.

Our survey team included soil technicians from the Lilongwe University of Agriculture and Natural Resources (LUANAR) soil lab. The technicians took two soil samples from each selected plot. Technicians took one sample from the topsoil (0–20 cm) and a second sample 20–40 cm below the surface (subsoil). Having a measure of quality for the topsoil and subsoil gave us a more robust estimate of the soil health on the plot.6

After collection and labeling, the LUANAR soil lab analyzed the samples for the following measures in the topsoil and subsoil: nitrogen (%), phosphorus (ppm), pH, organic matter (%), silt (%), clay (%), and sand (%). These quantitative measures allowed us to measure soil quality broadly. In addition, we asked all households about their perceptions of the soil quality on their owned, rented-in, and rented-out plots. This enabled us to compare quantitative measurements of soil quality with farmers’ assessments of soil quality. Furthermore, because we asked tenants and landlords about the soil quality on the rented plot that connected them, along with assessments of the soil quality on their respective owner-operated plots, we were able to observe how their perceptions about soil quality affected the land rental decision.

**Descriptive Statistics**

Table 1 presents the descriptive statistics comparing means and standard deviations of key covariates used in the analysis for tenants and their landlord pairs. It is clear from the table that the population of tenants was different from the population of landlords in our sample. The means of these variables were statistically different between tenants and landlords at the 1% level for all variables other than number of household members (0.05 < p < 0.10), if a member of the household belonged to a village savings and loan association (0.01 < p < 0.05), and walking distance to the nearest extension officer (p > 0.10). Table 1 shows that tenants had on average 3.10 more years of education than landlords. Tenant households also had 0.47 more

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6 Due to collinearity, we only used topsoil measures for phosphorous, soil organic matter, and pH. Other measures were dropped.
members on average than landlords, suggesting that tenant households had more available labor and more mouths to feed. Tenant household heads were also statistically younger than landlords on average by more than seven years and less likely to be female headed by 15 percentage points. Furthermore, tenants had a much higher average value of nonland assets at US$737 compared with US$117 for landlords, a difference of US$620. In fact, the only asset that landlords seemed to have statistically more of than tenants was prerental landholding, with the average landlord holding 1.88 ha, and the average tenant holding 0.83 ha. These statistics are consistent with other studies from southern Africa, which suggested that tenants on average have more nonland resources than do landlords (Chamberlin and Ricker-Gilbert 2016; Ricker-Gilbert et al. 2019). Our descriptive statistics provide prima facie evidence of tenants bringing education and assets into agriculture and using those resources to acquire land from less well-off landlord households.

Table 2 compares the means in input usage and soil fertility investments on tenants’ rented-in plots (column [1]), tenants’ owner-cultivated plots (column [2]), and landlords’ owner-cultivated plots (column [3]). First, we found that tenants were statistically more likely to apply purchased inputs—particularly herbicide and inorganic fertilizer—than were landlords, on average. However, tenants did not apply these inputs at statistically different rates for their rented-in plots (column [1]) versus their owner-cultivated plots (column [2]). However, longer-term investments showed strong statistical differences between tenant’s owner-operated and rented-in plots: rented-in plots received less animal manure, which is consistent with other empirical studies previously cited (e.g., Gavian and Fafchamps 1996; Jacoby and Mansuri 2008; Muraoka, Jin, and Jayne 2018). Although we do not know what investments landlords might have made on the rented-out plots had they not been rented out, it is nonetheless worth noting that in comparing the rented-in plots (operated by tenants) in column (1) with the landlord’s owner-operated plots in column (3), we found that there were statistically lower levels of longer-term investments on the rented-in plots managed by tenants for manure application and green compost application compared with the landlord’s owner-operated plots.

Considering the evidence from Table 2, that soil fertility investment seemed to be lower on rented-in plots than on owner-operated plots,
Table 3 provides some descriptive evidence on factors associated with which plots landlords decided to rent out versus cultivate themselves. The table shows objective measures and landlords’ subjective perceptions of the soil quality characteristics of their plots. Interestingly, most quality measures between the landlords owned versus rented-out plots were not statistically different. Landlords reported that their owner-operated and rented-out plots are of good quality at similar rates (45% and 47%, respectively), and objective measures of soil acidity, organic matter, and phosphorous were not statistically different from one another. There was no statistically significant difference in tenure insecurity on the plots that the average landlord rented-out versus cultivated themselves. However, rented-out plots were 17 percentage points less likely to have fruit trees planted on them (25% of plots) compared with landlords’ owner-operated plots (42% of plots) ($p < 0.01$). These findings suggest that the soil quality of plots did not strongly influence whether they were rented out, although the presence of fruit trees—as income- and sustenance-generating assets—was likely important.

5. Input Use and Soil Fertility Investments

This article seeks to provide insights into the relationships between soil fertility, land rental markets, and tenant and landlord interactions in Malawi. In this section we discuss how we test the key research questions presented in the article.

Question 1: How did input use and soil fertility investment decisions differ across owned versus rented plots operated by tenants and landlords? First, we want to estimate how land rental markets affect input use and soil fertility investments. The specific plot-level inputs that we model in the study are (1) the number of times the plot was weeded, (2) if the farmer applied herbicides on the plot, (3) kilograms of inorganic fertilizer applied per hectare to the plot, and (4) if the farmer planted hybrid maize as the main crop on the plot. We would expect all of these to increase land productivity (yields) in the current year, but they all come with extra costs over traditional low-input farming systems. We briefly discuss these variables and the associated benefits and costs below.

Number of Times the Plot was Weeded

The dependent variable number of times the plot was weeded is an important input on the plot because effective weed management is widely known to encourage crop growth and enhance the nitrogen fertilizer use efficiency (Snapp et al. 2014). At the same time, weeding incurs labor costs, which affect the optimal number of times a plot is weeded. For example, Kamanga et al. (2014) found that, on average, when inorganic fertilizer was valued at highly subsidized prices through...
the Farm Input Support Program in Malawi, the financial returns to labor were generally higher when farmers weed twice instead of once. Conversely, when they valued fertilizer at commercial prices, they found that weeding once or twice did not generate returns to labor that exceeded the prevailing informal hired-in (ganyu) labor wage rate of $0.53 per day. Appendix Table A2 shows that in our data the median wage rate per day (four hours of work) for ganyu labor during the 2015–2016 season in Malawi was MWK 528, roughly equivalent to US$0.75, and that 45% of the sample hired in some ganyu labor. In addition, 50% of the plots in our data were weeded once and 41% were weeded twice.

**Applied Herbicides**

Herbicides are a technology to control weeds that can help farmers maintain yields and reduce labor costs, compared to weeding by hand in traditional smallholder cultivation systems. Furthermore, applying herbicides are an important input in minimum or zero tillage methods that are recommended as sustainable agricultural practices that farmers are often encouraged to adopt (Giller et al. 2009, discussed more below). The challenge for many smallholders is that applying herbicides means that they have to incur an extra input cost, and access to herbicides is severely limited in many parts of Malawi. Table 2 indicates that 21% of tenant’s rented-in plots had herbicides applied to them, 24% of tenants’ owner-cultivated plots had herbicides applied, and just 9% of landlords’ owner-cultivated plots had herbicides applied. We did not attempt to value or create a price for herbicide because of the difficulty in standardizing the different types and quantities that farmers use. Regardless, the vast majority of smallholders do not apply herbicide regardless of ownership or cultivation status and have to resort to using traditional hand weeding methods.

**Applied Inorganic Fertilizer**

Inorganic fertilizer is regarded as an important yield-increasing input in maize production in Malawi’s nitrogen-deficient soils (Maize Productivity Task Force, Action Group I 1999). The government of Malawi recommends that farmers apply inorganic fertilizer two times to their maize. The most common fertilizer blends in Malawi are NPK, used as basal fertilizer applied during planting, and urea used as a top dressing to be applied once maize has sprouted. Appendix Table A2 indicates that 65% of our sample applied some amount of inorganic fertilizer during 2015–2016. The median farmer in the sample who used inorganic fertilizer applied 100 kg, which is the recommended amount for one acre. The median commercial fertilizer price was MWK 360/kg roughly equivalent to US$0.50. This costs out to about US$50 in total expenditure on inorganic fertilizer at commercial prices for people who purchase it in Malawi. This is a significant financial outlay for many smallholders.

**Planted Hybrid Maize**

Hybrid maize varieties have the potential to increase yields compared with traditional varieties. Hybrid varieties are shorter in stature than traditional varieties, and more of the plant’s energy is transferred into the grain rather than the stock of the plant. In addition, many of the hybrid varieties in Malawi are early maturing and drought resistant, allowing farmers to maintain yields in response to more sporadic rainfall and drought that has occurred in recent years, including during 2015–2016 when our data were collected. Snapp et al. (2014) found that on average yields on hybrid maize plots in Malawi were 1,373 kg/ha and 1,289 kg/ha on plots with local maize varieties.

Although there are yield benefits to hybrid seeds, there are additional costs and other potential drawbacks. First, hybrid seeds lose productivity if they are recycled, and seed should be purchased every year, compared with traditional varieties that can be saved and reused in multiple seasons. Second, Malawians tend to prefer the taste of traditional maize grain and traditional maize grain stores better than hybrid maize, so the former often fetches a higher output price at market (Lunduka, Fisher, and Snapp 2012). Appendix Table A2 indicates that 65% of households in our sample acquired hybrid maize seed through the subsidy program or by purchasing them on
the commercial market. The table also shows that the median farmer who purchased commercial hybrid maize seed bought 10 kg, recommended for planting one acre. The median cost of commercial maize seed was MWK 700/kg, roughly equal to US$1.00.

The plot-level soil fertility investments modeled in this article include (a) intercropping maize and legumes, (b) applying animal manure, (c) applying green compost, and (d) using minimum tillage methods. These practices can maintain or enhance soil fertility, leading to increased yields, profits, and potentially better nutrition and food security. However, when implementing these practices, it takes time for the benefits to be translated into increased productivity. This is especially true with minimum tillage. These practices impose additional costs in terms of labor and the need for complimentary input use (e.g., herbicides to control weeds when using minimum tillage). We briefly discuss these soil fertility–enhancing investments below.

Intercropped Maize and Legumes

The general principle of intercropping in smallholder production systems involves planting maize with legumes, including common bean, groundnuts, pigeon pea, or soybean. Legumes fix nitrogen in the soil, which offsets the nitrogen that maize takes out of it. Over time this can reduce nitrogen depletion and reduces (but certainly does not eliminate) the need for inorganic fertilizer. Bell et al. (2018) surveyed farmers in Malawi’s Shire River valley and found that the main reasons they intercropped maize with legumes were to diversify their crop portfolio, reduce risk of maize crop failure, and have more (nutritious) food to eat and sell. The potential drawbacks to maize and legume intercropping are that maize production intensity and yields decline with lower planting densities, and planting multiple crops together on a plot increases the amount of labor needed. The other challenge is that legume seeds have to be available when farmers need them at planting, which can be a problem for smallholders in Malawi. Kanyamuka, Jumbe, and Ricker-Gilbert (2018) found that the main reason adopters of the legume pigeon pea in Central Malawi stopped growing the crop was because they did not have consistent access to fresh seed.

Applied Animal Manure

Applying organic animal manure to plots is generally viewed as a sustainable agricultural intensification strategy that can be used as an agronomic compliment to inorganic fertilizers in smallholder production systems. Manure adds organic matter to the soil and improves its structure so that plants can use additional nitrogen from inorganic fertilizer more efficiently. However, it takes time for organic matter to accumulate in the soil and for soil structures to improve (Holden and Lunduka 2012). At the same time, organic manure is bulky and expensive to transport compared with inorganic fertilizer. Due to severe land constraints in Malawi, particularly in the southern part of the country, livestock density is low, making the benefits of organic manure difficult to access for limited resource smallholders.

Applied Green Compost

Green compost is created from crop residues, branches, and other greenery that smallholders apply to their fields to build up organic matter in their soils. Mulching green compost into the soil can improve its fertility over time, but like animal manure, it takes several years for the benefits to be realized (Van Hulst and Posthumus 2016). In addition, crop residues have competing uses such as livestock feed, composting takes time to create, and it is labor intensive and bulky to transport and spread (Giller et al. 2009).

Used Minimum Tillage

Minimum or zero tillage systems where farmers avoid creating deep disturbances in the soil is often advanced as being an important component of sustainable agricultural intensification strategies (Giller et al. 2009). Minimum tillage is also considered as a key component of conservation agriculture (Ngoma et al. 2021). Leaving the soil untouched rather than creating ridges saves labor and increases fertility of the soil over time. It can also help
prevent erosion and water logging caused by poorly constructed ridges. However, ridging is a traditional weed control strategy in smallholder production systems, and minimum tillage is not effective unless farmers have access to herbicides to control weeds (Ward et al. 2018). Given the resource constraints of many smallholder farmers and the lack of access to herbicides, it is difficult for many of them to practice minimum tillage methods.

6. Empirical Models

Our first empirical model considers how these input use and soil fertility–enhancing investments differ among plots with three distinct types of ownership and cultivation status: (A) a tenant’s rented-in plot that he or she cultivates, (B) the tenant’s owner-operated plot, and (C) the landlord’s owner-operated plot. We model these relationships on plot cultivated by household j in rental-pair p as follows:

\[
X_{ijp} = \beta_1 R_{ijp} + \beta_2 T_{ijp} + H_{jp} \beta_3 + \beta_4 D_{ijp} + c_p + \epsilon_{ijp},[1]
\]

where \( X \) represents an input or soil fertility investment decision on the plot. The variable \( R \) is a binary indicator, equal to one if the plot belongs to plot status A, namely, that it is rented-in and thus cultivated by the tenant, while \( T \) represents the corresponding parameter to estimate. The variable \( T \) is also a binary indicator, equal to one if the plot belongs to plot status B, the plot is owner-operated by the tenant. The corresponding parameter to estimate is \( \beta_2 \). The interpretation of the coefficient estimates \( \beta_1 \) and \( \beta_2 \) is important for answering how input use and investments differ across the types of plot ownership and cultivation status. The sign and magnitude of \( \beta_1 \) tests and compares how input use and investments differ on a tenant’s rented-in plot (plot status A) compared to his or her landlord’s owner-operated plot (plot status C). The sign and magnitude of \( \beta_2 \) tests and compares how input use and investments differ on a tenant’s owner-operated plot and his or her landlord’s owner-operated plot (status B versus status C). We can compare the differences between a tenant’s rented-in plot and his or her owner-operated plot (status A versus status B) with an F-test of equality between \( \beta_1 \) and \( \beta_2 \).

Given the short-term nature of rental contracts in Malawi, we might expect input used to be higher among tenants and on rented-in plots compared with landlord’s owner-operated plots because the tenant operators seek short-term yield gains on the rented land. As mentioned, previous literature from Malawi indicated that tenants were also likely to have more savings and assets than landlords, so were likely more able to purchase fertilizer, seed, pesticides, and other inputs. At the same time, we hypothesize that soil fertility investments might be lower on rented-in plots than they are on owner-cultivated plots, either managed by tenants or landlords. This is because of the longer time horizon for these benefits to materialize and the likely shorter-term nature of the rental arrangements, as seen in previous literature (Gavian and Fafchamps 1996; Muraoka, Jin, and Jayne 2018). The comparison of the coefficient estimates \( \beta_1 \) and \( \beta_2 \) to each other and to the control (landlord’s owner-operated plots) gives a more complete picture of decisions made by land owners and land operators that has yet to be analyzed (to our knowledge).

Equation [1] also includes a set of household-level controls that are denoted by the vector \( H \). These include prerental landholding, which includes all land cultivated by the household (excluding rented-in land) in addition to land that will be rented out, and land that is fallowed, used as a woodlot or in pasture. The vector \( H \) also includes number of household members, a proxy for available family labor, along with number of mouths to feed. In addition, gender of the household head, age of the household head, household savings, value of household assets, number of plots cultivated by the household, if someone the household is a member of a village savings

---

7 We note that this variation is only useful in this context if renters have multiple plots with differing rental status. In our data set, we have very few households with only a single plot. Summary statistics are provided in Appendix Table A1.

8 Land may also be borrowed-in, where one household lets another household cultivate their land with no money exchanged. For the purpose of this analysis, we consider borrowed land to be rented land at a zero price.
and loan association, and walking distance to nearest extension office are all included as control variables in $H$. The corresponding parameter vector in equation [1] is denoted by $\beta_3$. We also add one other plot-level variable: distance to the plot from the operator’s home in walking minutes. This factor is denoted by $D$, with corresponding parameter $\beta_4$.

The rental-pair-specific unobserved FE that could influence the rental decision is denoted by $c$ in equation [1]. This variable captures unobserved differences across tenant-landlord pairs that could influence unobservable factors, such as ability, motivation, and which plots have been rented in or out. Such unobservable variables include social and power dynamics and social connections within the rental partner pair (Bellemare 2012; Deininger, Ali, and Alemu 2013). We deal with this potential source of endogeneity by estimating equation [1] using rental-pair FE. Rental-pair FE controls for unobserved heterogeneity in the tenant-landlord pair, but as Bellemare (2012) pointed out, it does not control for selection into rental status as a tenant or landlord. Rather, the results of this analysis are consistent for measuring effects conditional on being engaged in the land rental market as a tenant or landlord. The results are not fully generalizable to the broader population of smallholders in Malawi but are relevant for people who are already either tenants or landlords. We argue that this is the primary population of interest here because some people in the smallholder population are unlikely to ever engage in land renting. Ultimately, we recognize that even with a rich set of controls and pair-specific FEs, we cannot assume full causality of our results. This is the case with any study using observational data, but we believe our analysis uncovers important relationships that are useful for smallholder agricultural policy in SSA.

The plot and household specific error term is represented by $\epsilon$, which is assumed to be uncorrelated with the observable covariates in equation 1 after controlling for rental-pair FE.

Question 2: How do subjective and objective measures of soil fertility along with investments affect a landlord’s rental decision? To answer this question, we explicitly model the landlord household j’s rental decision for plot i at the beginning of the agricultural season as a function of the following:

$$Y_{ij} = a_1 S_{ij} + a_2 I_{ij} + Q_{ij} a_3 + a_4 C_{ij} + a_j + v_{ij},$$

where $Y$ is equal to one if the landlord chooses to rent out the plot and zero if she chooses to cultivate it herself. The first key covariate to consider in our analysis is denoted by $S$, which is equal to one if the household believes that the soil fertility on plot $i$ is of good or very good quality and zero otherwise. The corresponding parameter to estimate is denoted by $a_1$. The sign and statistical significance tests whether a landlord rents out the plot that she perceives to be of better or worse soil fertility.

The variable $I$ in equation [2] represents the binary indicator for whether there is a fruit tree on the plot. This variable is meant to proxy for the investment decisions made in previous years on the plot that may affect household income or the plot’s soil fertility in the future. The fruit that the trees bear serves as a resource that the household can consume or sell, with $a_2$ as the corresponding parameter. This variable tests whether landlords are more or less likely to rent out the plots that have received an investment previously. Appendix Table A2 shows that 58% of households have ever purchased a fruit tree; conditional on purchasing a tree, five was the median number of trees purchased and six was the mean number purchased. The median fruit tree value was MWK 2,000 per tree (US$2.85), while the mean was MWK 3,779 (US$5.40).

One may wonder if a fruit tree on a rented plot could have been planted by a tenant in the past rather than by the landlord. There is no way to rule this out for certain. However, it seems unlikely because most of the renting in this context is very short term, as evidenced in the previous literature (Lunduka, Holden, and Øygard 2009; Chamberlin and Ricker-Gilbert 2016; Ricker-Gilbert et al. 2019). Furthermore, Appendix Table A3 shows that 78% of rented plots in our data were rented out for two years or less over the past five years. Because fruit trees are a relatively fixed investment, this suggests that any trees planted on
rented-out plots were likely planted by the owning landlord sometime in the past.\(^9\)

The vector \(Q\) represents the unobservable, quantitative soil fertility measures on the plot’s top soil (0–20 cm). These include the level of phosphorus measured in parts per million (ppm), level of organic matter in percent, and if the soil is acidic with pH below 5.2. The corresponding parameter vector is \(\alpha_3\). These are important measures for testing the extent to which landlords are aware of objective soil fertility on the plot and how this may affect their rental decisions.

We also include the variable \(C\) in equation [2] to provide us with a measure of how tenure insecurity affects the decision to rent. The variable is equal to one if the landlord believed that someone was likely to make a tenure claim against his or her plot. The corresponding parameter is \(\alpha_4\). The error term in equation [2] has two components. First, the landlord-specific unobserved effect (FE) is denoted by \(a\), while \(v\) denotes the household and plot-specific error.

The identification strategy used to deal with correlation between the error term and observed covariates in equation [2] is similar to equation [1], as we are primarily concerned with omitted variable bias. We deal with this first by adding the plot-level controls as mentioned above, thus removing them as omitted variables in the model.

Second, because we have multiple plots per household, we are able to use landlord FE to remove the individual landlord-specific error term, \(a\), from equation [1]. Doing so removes demographic characteristics of the landlord, such as age, education, and gender, as those are constant for an individual landlord. The plot-specific error, \(v\), is assumed to be uncorrelated with the covariates in equation [2], conditional on the observed covariates and \(a\).

### Estimator Choice

Most of the dependent variables in the models presented in this article are binary responses, taking on a zero or one value (the only exceptions are number of times the plot was weeded and kilograms of inorganic fertilizer applied per hectare). We estimate these linearly using rental-pair FE as discussed earlier. As such, the binary dependent variables in these equations are estimated as an LPM. LPM has the advantage over a nonlinear estimator such as probit because it provides easy to interpret coefficients. LPM also allows us to use tenant-landlord pair FE, which would be biased in probit estimation, due to the incidental parameters problem (Wooldridge 2010). We cluster the standard errors of our estimates at the rental-pair level to deal with concerns about heteroskedasticity and serial correlation when using the LPM.

### 7. Results

Table 4 presents the results for equation [1] that estimates the factors affecting plot-level input use among (1) a tenant’s rented-in plot(s) \((\hat{\beta}_1)\), (2) a tenant’s owner-operated plot(s) \((\hat{\beta}_2)\), and (3) his or her landlord’s owner-operated plot(s) as the base for comparison. These models were estimated linearly via rental-pair FE to control for unobservable factors in the relationship between tenants and landlord pairs. We see that there was no statistical difference among ownership status and management of the plot for number of times the plot was weeded in column (1). However, in column (2), tenants were 8 percentage points more likely to apply herbicide on their owner-operated plot than were landlords on their owner-operated plot \((p < 0.10)\). This was nearly double the 9% of landlords who applied herbicides to their owner-operated plots (as seen in Table 2). The same relationship held in column (3) for kilograms of inorganic fertilizer applied per hectare. Tenants applied 78 kg per ha more fertilizer on average on plots they owner-operated, compared with their landlords’ owner-operated plots \((p < 0.01)\). This was a substantial 51% increase over the mean fertilizer application on a landlord’s plot of 152 kg per ha (as seen in Table 2). Tenants also applied nearly 61 kg more inorganic fertilizer per ha on rented-in plots than their landlords did on their owner-operated plots on average \((p < 0.05)\). This was equivalent to a 40% increase in fertilizer application over their application.
landlord’s owner-operated plot(s) on average. However, we did not observe statistically significant differences between tenant-operated plots (both rented-in or owner-operated) and landlords’ owner-operated plots in terms of number of times the plot was weeded or whether hybrid maize was the main crop.

To evaluate differences between input usage by tenants on their owned versus rented-in plots, we computed $F$-tests comparing $\hat{\beta}_1$ and $\hat{\beta}_2$ for each model. The only significant difference was in column (4) (hybrid maize usage): tenants were 11 percentage points more likely to plant hybrid maize on their owned-plot than on their rented-in plots on average. (Appendix Table A4 shows the main crops that were planted on plots based on ownership and cultivation status.) One revealing insight from Appendix Table A4 is that tenants grew groundnuts on nearly 30% of rented-in plots, compared with groundnuts being grown on only 17% of tenants’ owner-cultivated plots and 19% of landlords’ plots. This may suggest that tenants focused on planting maize as a food security crop on their owned plots but were more likely to use the rented-in plots to expand their area cultivated into cash crops like groundnuts.

Table 4
Factors Affecting Plot-Level Input Use

<table>
<thead>
<tr>
<th>Variable</th>
<th>(1) No. Times Plot Was Weeded</th>
<th>(2) $1=\text{Applied Herbicide}$</th>
<th>(3) Inorganic Fertilizer Applied (kg/ha)</th>
<th>(4) $1=\text{Hybrid Maize as Main Crop}$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\hat{\beta}_1$</td>
<td>$\hat{\beta}_2$</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>$-$0.04</td>
<td>$-$0.03</td>
<td>$61.01^{**}$</td>
<td>$-$0.03</td>
</tr>
<tr>
<td></td>
<td>$(0.131)$</td>
<td>$(0.116)$</td>
<td>$(29.107)$</td>
<td>$(0.059)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.00</td>
<td>$-$0.01**</td>
<td>$78.07^{***}$</td>
<td>$0.08$</td>
</tr>
<tr>
<td></td>
<td>$(0.010)$</td>
<td>$(0.005)$</td>
<td>$(27.846)$</td>
<td>$(0.061)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.00</td>
<td>$-$0.01**</td>
<td>$3.04$</td>
<td>$0.01^{*}$</td>
</tr>
<tr>
<td></td>
<td>$(0.010)$</td>
<td>$(0.016)$</td>
<td>$(2.624)$</td>
<td>$(0.006)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.01</td>
<td>$-$0.01**</td>
<td>$-$1.81</td>
<td>$0.01$</td>
</tr>
<tr>
<td></td>
<td>$(0.003)$</td>
<td>$(0.016)$</td>
<td>$(11.136)$</td>
<td>$(0.021)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.01</td>
<td>$-$0.01**</td>
<td>$-$14.91^{***}</td>
<td>$-$0.01</td>
</tr>
<tr>
<td></td>
<td>$(0.022)$</td>
<td>$(0.009)$</td>
<td>$(5.402)$</td>
<td>$(0.013)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.04</td>
<td>$-$0.01**</td>
<td>$0.33$</td>
<td>$0.00$</td>
</tr>
<tr>
<td></td>
<td>$(0.004)$</td>
<td>$(0.002)$</td>
<td>$(0.760)$</td>
<td>$(0.002)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.00</td>
<td>$-$0.01**</td>
<td>$0.01$</td>
<td>$0.00$</td>
</tr>
<tr>
<td></td>
<td>$(0.000)$</td>
<td>$(0.000)$</td>
<td>$(0.030)$</td>
<td>$(0.000)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.02</td>
<td>$-$0.03**</td>
<td>$2.31$</td>
<td>$0.01$</td>
</tr>
<tr>
<td></td>
<td>$(0.025)$</td>
<td>$(0.014)$</td>
<td>$(6.455)$</td>
<td>$(0.014)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.00</td>
<td>$-$0.00 **</td>
<td>$0.02$</td>
<td>$-$0.00</td>
</tr>
<tr>
<td></td>
<td>$(0.001)$</td>
<td>$(0.001)$</td>
<td>$(0.517)$</td>
<td>$(0.001)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.00</td>
<td>$-$0.00 **</td>
<td>$-$17.05</td>
<td>$-$0.11^{**}</td>
</tr>
<tr>
<td></td>
<td>$(0.000)$</td>
<td>$(0.000)$</td>
<td>$(0.099)$</td>
<td>$(0.000)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.01</td>
<td>$-$0.02**</td>
<td>$-$17.05</td>
<td>$-$0.11^{**}</td>
</tr>
<tr>
<td></td>
<td>$(0.074)$</td>
<td>$(0.026)$</td>
<td>$(0.371)$</td>
<td>$(0.040)$</td>
</tr>
<tr>
<td></td>
<td>$-$0.01</td>
<td>$-$0.02**</td>
<td>$-$17.05</td>
<td>$-$0.11^{**}</td>
</tr>
<tr>
<td></td>
<td>$(0.074)$</td>
<td>$(0.026)$</td>
<td>$(0.371)$</td>
<td>$(0.040)$</td>
</tr>
</tbody>
</table>

Note: The base category for comparing $\hat{\beta}$ estimates is a landlord’s owner-operated plot(s). Standard errors are in parentheses. Models include a constant and district-level FE. Districts are not completely collinear with the rental-pair location because we found four cases of tenants and landlord pairs residing in different districts. The number of observations = 948, with 169 matched tenant-landlord pairs; $\hat{\beta}_1$ and $\hat{\beta}_2$ are compared with the landlord’s owner-cultivated plot. ***, **, * Significance at the 1%, 5%, and 10% levels, respectively.
Table 5 presents the results for factors affecting plot-level soil fertility investments among a tenant’s rented-in plot(s) ($\hat{\beta}_1$) a tenant’s owner-operated plot(s) ($\hat{\beta}_2$), and his or her landlord’s owner-operated plot(s) (control). As in Table 4, the models in Table 5 were estimated linearly via rental-pair FE. For the most part, there are no statistically significant differences in the probability of longer-term investments for tenants’ rented-in and owner-operated plots compared with their landlords’ owner-cultivated plots. The only significant exception is that, on average, rented-in plots were 7 percentage points less likely to receive green compost by a tenant operator, compared with a landlord’s green compost usage on their owner-operated fields. This result makes sense given the time, labor, and crop residues required to generate green compost: because crop residues have alternative uses and land renting is short-term, tenants may have greater incentives to apply such resources on their own plots, where lagged benefits will not be lost.

<table>
<thead>
<tr>
<th>Variable</th>
<th>(1) Intercropped Maize and Legume</th>
<th>(2) Applied Animal Manure</th>
<th>(3) Applied Green Compost</th>
<th>(4) Used Minimum Tillage</th>
</tr>
</thead>
<tbody>
<tr>
<td>$1$ if plot rented in and cultivated by tenant $\hat{\beta}_1$</td>
<td>0.04 (0.048)</td>
<td>-0.05 (0.049)</td>
<td>-0.07** (0.034)</td>
<td>-0.02 (0.051)</td>
</tr>
<tr>
<td>$1$ if plot owned and cultivated by tenant $\hat{\beta}_2$</td>
<td>0.08 (0.047)</td>
<td>0.06 (0.054)</td>
<td>-0.02 (0.038)</td>
<td>0.03 (0.051)</td>
</tr>
<tr>
<td>Education of household head in years of schooling</td>
<td>-0.00 (0.005)</td>
<td>-0.01 (0.006)</td>
<td>0.00 (0.004)</td>
<td>-0.00 (0.006)</td>
</tr>
<tr>
<td>Area owned by household pre land-renting, in hectares</td>
<td>-0.02 (0.020)</td>
<td>0.01 (0.021)</td>
<td>-0.02 (0.012)</td>
<td>0.03 (0.024)</td>
</tr>
<tr>
<td>Number of members in household</td>
<td>0.02** (0.009)</td>
<td>-0.02** (0.009)</td>
<td>-0.01 (0.007)</td>
<td>0.01 (0.010)</td>
</tr>
<tr>
<td>$1$ if household head is female</td>
<td>0.06 (0.063)</td>
<td>-0.10* (0.059)</td>
<td>-0.00 (0.034)</td>
<td>0.01 (0.046)</td>
</tr>
<tr>
<td>Age of household head in years</td>
<td>0.00 (0.001)</td>
<td>0.00 (0.002)</td>
<td>0.00* (0.001)</td>
<td>0.00 (0.001)</td>
</tr>
<tr>
<td>Savings in USD*1,000</td>
<td>0.00* (0.000)</td>
<td>0.002* (0.000)</td>
<td>0.00 (0.000)</td>
<td>0.00 (0.000)</td>
</tr>
<tr>
<td>Value of assets in USD*1,000</td>
<td>-0.00 (0.000)</td>
<td>-0.00 (0.000)</td>
<td>-0.008*** (0.000)</td>
<td>-0.01*** (0.000)</td>
</tr>
<tr>
<td>Number of plots cultivated by household</td>
<td>-0.05*** (0.013)</td>
<td>-0.01 (0.018)</td>
<td>0.00 (0.009)</td>
<td>-0.01 (0.013)</td>
</tr>
<tr>
<td>Plot distance from house (walking minutes)</td>
<td>0.00 (0.001)</td>
<td>-0.00 (0.001)</td>
<td>0.00 (0.001)</td>
<td>0.00 (0.001)</td>
</tr>
<tr>
<td>$1$ if member of household is in a village savings and loan association</td>
<td>0.07* (0.042)</td>
<td>0.01 (0.057)</td>
<td>0.03 (0.042)</td>
<td>0.03 (0.036)</td>
</tr>
<tr>
<td>Distance to the nearest ag. extension officer from residence (walking minutes)</td>
<td>0.0002* (0.000)</td>
<td>0.00 (0.000)</td>
<td>0.00 (0.000)</td>
<td>0.00 (0.000)</td>
</tr>
<tr>
<td>$\hat{\beta}_1 - \hat{\beta}_2 = 0$</td>
<td>-0.04 (0.028)</td>
<td>-0.11*** (0.034)</td>
<td>-0.05** (0.022)</td>
<td>-0.05* (0.026)</td>
</tr>
<tr>
<td>Rental-pair fixed effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>$R$-squared</td>
<td>0.046</td>
<td>0.041</td>
<td>0.038</td>
<td>0.046</td>
</tr>
<tr>
<td>Observations</td>
<td>948</td>
<td>948</td>
<td>948</td>
<td>948</td>
</tr>
</tbody>
</table>

Note: Standard errors are in parentheses. Models include a constant and district-level FE. Districts are not completely collinear with the rental-pair location because we found four cases of tenants and landlord pairs residing in different districts. The number of observations = 948, with 169 matched tenant-landlord pairs; $\hat{\beta}_1$ and $\hat{\beta}_2$ are compared with the landlord’s owner-cultivated plot. ***, **, * Significance at the 1%, 5%, and 10% levels, respectively.
indicated that tenants were much less likely to use animal manure, green compost, and minimum tillage practices on their rented-in plots than they were on their owner-operated plots by 11, 5, and 5 percentage points on average, respectively. These changes were very different from the means on tenant’s owner-cultivated plots, representing a 42% decline in the probability of applying animal manure, a 45% decline in the probability of using green compost, and a 35% decline in the probability of using green compost, albeit from a relatively low base (as seen in Table 2).

Regardless, these results were all in line with our expectations, and consistent with the previous literature about the inability of tenants to capture lagged returns under short-term tenancy arrangements (Gavian and Fafchamps 1996; Jacoby and Mansuri 2008; Lovo 2016; Muraoka, Jin, and Jayne 2018). As mentioned, these soil fertility-enhancing investments take time to generate returns. Furthermore, organic animal manure requires access to animals and labor; green compost requires access to labor and crop residues, which have alternative uses; and minimum tillage is enabled by herbicide application, which is an added input cost and access is very limited for many smallholders in Malawi. Obviously, if tenants have to make decisions about where to incur these costs to obtain soil fertility benefits from these investments, they generally choose to do so on their own plots rather than the rented ones. As such, it would seem that tenants in our sample were not investing in their rented-in plots and may in fact have been mining the soil nutrients on these plots. We also know that tenants in our sample were wealthier than landlords on average in terms of wealth, savings, and education. These findings raise the question that we address in Table 6: what factors affected the landlord’s choice of which plot to rent out and which to retain for her own cultivation?

Table 6 presents the results of the model that estimated factors associated with which plot(s) a landlord decided to rent out, estimated via landlord-FE LPM. Columns (1)–(4) show the base model presented in equation [2] with alternative sets of explanatory factors in each column, while the specification in column (5) includes all factors simultaneously. We see from these results that a landlord’s perception of the soil quality on the plot did not significantly affect his or her decision to rent out the plot. However, in columns (3) and (5), we see that plots with fruit trees on them were 49–51 percentage points less likely to be rented out, on average $(p < 0.01)$. This suggests that landlords viewed fruit trees as important productive assets and were reluctant to part with plots that had such trees.

We also found evidence in column (2) that plots with higher soil organic matter were less likely to be rented out on average, with results approaching statistical significance $(0.10 < p < 0.11)$.

Interestingly, we found marginally significant evidence that landlords who felt insecure about their tenure claim to a plot were 29–33 percentage points more likely to rent it out on average than those who did not $(0.05 < p < 0.10$ in column [5]). One possible interpretation of this result is that if landlords were labor or capital constrained, and unable to farm all the land in their possession, they may have perceived a greater risk of appropriation/reallocation from leaving the land idle rather than letting a tenant cultivate it. Another possibility is that, facing some uncertainty about eventual reallocation by traditional authorities, a landlord would rather earn...
We investigated this finding further by estimating a plot-level model of factors affecting the landlord’s perception of insecurity across their plots. The model regressed a binary variable for insecurity (i.e., the expressed concern by the respondent that someone will challenge his or her tenure claim) on (1) distance from the home to the plot, (2) his or her subjective views on soil quality and soil color, (3) if the plot was obtained from the local chief (as opposed to inheriting it directly from a parent, or purchasing it with title), and (4) if the male in the household was the main operator of the plot. The estimation results of this model are shown in Appendix Table A5. They indicate that obtaining the plot from a chief had a marginally significant effect on landlord’s perception of insecurity about the plot at the 10% level of significance. In fact, the coefficient estimate suggests that the average household who obtained the plot from the chief was 29%–31% more likely to feel insecure about their claim to it. This finding suggests that ceteris paribus when smallholders obtain plots directly from chiefs, rather than through inheritance claims or direct purchases, they are more likely to perceive a risk of having that land taken away. This may reflect that land obtained directly from chiefs was probably acquired more recently and could be more easily taken away, relative to land that has been inherited and controlled by the farmer’s family for multiple generations.

Robustness Checks

For each model specification results reported in Tables 4–6, we also generated a number of alternative specifications, with the objective of evaluating the robustness of our results. In Appendix Tables A6 and A7, we ran parsimonious models with and without pair-wise FE, as well as a specification with a full set of covariates but without the pair-wise FE. In Appendix Table A8 we show specifications that were analogous to the main landlord results in Table 6, but without landlord FE. Across the specifications in the appendixes, the signs were all the same and coefficients were generally of similar magnitude to our main results. Fewer coefficients were statistically significant in Tables 4–6 than they were in the appendixes, suggesting that the FE estimators with additional controls provided the

### Table 6

Factors Affecting the Plot the Landlord Decides to Rent Out

<table>
<thead>
<tr>
<th>Dependent Variable (=1 if plot rented out, =0 if owner cultivated)</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
</tr>
</thead>
<tbody>
<tr>
<td>=1 if landlord views soil as good or very good</td>
<td>0.11</td>
<td>0.02</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>in topsoil (&lt;5.2 pH)</td>
<td></td>
<td></td>
<td>0.05</td>
<td>0.04</td>
<td></td>
</tr>
<tr>
<td>Organic matter (%) in topsoil</td>
<td></td>
<td></td>
<td>0.07</td>
<td>0.06</td>
<td></td>
</tr>
<tr>
<td>Phosphorus (ppm) in topsoil</td>
<td></td>
<td></td>
<td>0.00</td>
<td>0.00</td>
<td></td>
</tr>
<tr>
<td>Fruit trees on the plot</td>
<td></td>
<td></td>
<td>0.49**</td>
<td>0.51**</td>
<td></td>
</tr>
<tr>
<td>=1 if landlord perceives someone likely to challenge tenure</td>
<td></td>
<td></td>
<td>0.29</td>
<td>0.33*</td>
<td></td>
</tr>
<tr>
<td>Landlord fixed effects</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Yes</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.004</td>
<td>0.018</td>
<td>0.097</td>
<td>0.016</td>
<td>0.137</td>
</tr>
<tr>
<td>Observations</td>
<td>245</td>
<td>245</td>
<td>245</td>
<td>245</td>
<td>245</td>
</tr>
</tbody>
</table>

Note: Standard errors are in parentheses. Models include a constant and district-level FE. Districts are not completely collinear with the rental-pair location because we found four cases of tenants and landlord pairs residing in different districts. The number of observations = 245, with 137 owner-operated plots and 108 rented-out plots.

***, **, * Significance at the 1%, 5%, and 10% levels, respectively.
most conservative estimates, justifying why we used them in our main analysis. Overall, our analytical conclusions did not change substantively when we added these robustness checks.

8. Discussion

Our findings are relevant to the contemporary policy debate on how best to facilitate sustainable intensification in SSA (Sanchez 2002; van Ittersum et al. 2016; Holden 2018; Jayne et al. 2019). Results suggest that the answer to this question is not simply a technological one but one that addresses how the incentives for soil fertility investments are conditioned by land institutions, including farmland rental markets. Population is growing rapidly in the region against a relatively fixed land base, and land rental markets are responding. Such markets are the most feasible way for land resources to be quickly and efficiently re-allocated to those who wish to expand their cultivated area and to invest management and financial resources in the sector. However, our results suggest that in the absence of other interventions, the de facto expansion of rental markets in SSA will likely incur negative effects on soil fertility, even if sectoral productivity increases in the shorter term.

This has implications for the long-run fertility and health of African soil, which are already in crisis (Drechsel et al. 2001; Sanchez 2002). The answer, surely, is not to restrict rental market development, which does appear to bring about important positive productivity gains (Holden, Otsuka, and Place 2009; Jin and Jayne 2013; Chamberlin and Ricker-Gilbert 2016), but to facilitate tenure security for landlords, which may facilitate longer-term rental arrangements, creating greater incentives for fertility investments by tenants. In addition, it may be possible to design contracts that incentivize longer-term investments directly. For example, in a study of land rental markets in the Czech Republic, Sklenicka et al. (2015) found that while owner-operators were more likely than tenants to invest in soil erosion control, tenants could be incentivized through subsidies to do so if an effective verification and payment system was put in place. This could perhaps be feasible in our context as we found evidence that tenants were more likely to invest in the soil fertility for their owner-operated plots than for their rented plots. This suggests that they know the benefits of soil fertility-enhancing measures but are not incentivized to care about them on their rented-in plots. Additional research may clarify the responsiveness of rental arrangements (and investments made in such rental contexts) to different levels of tenure security and contract choices.

Nonetheless, on the basis of our results, we consider some simple benefit-cost calculations on the basis of expected input intensification by landlords in the short-run versus the costs of decreased soil fertility in the longer run. Column (3) in Table 4 indicated that on average, moving land from landlords to tenants added an extra 61 kg per ha to Malawi’s soil. If we assume a maize to fertilizer response rate of 4:1 (Jayne and Rashid 2013), we obtain a return of an extra 244 kg of maize per ha compared with land managed by landlords. The nationally representative Fourth Integrated Household Survey of Malawi indicates that 13% of plots and 385,000 ha of land were rented during the 2015–2016 season. This suggests that an extra 23,500 tons of fertilizer was added to Malawi’s soils through land renting, leading to an additional 94,000 tons of maize produced (assuming a 4:1 fertilizer to maize response ratio). If we assume a maize grain price at harvest of $250 per ton, then the total revenue from the additional production would be US$23.5 million in 2015–2016.

Next the benefits of increased fertilizer and maize production need to be assessed against the cost of fertilizer and labor requirements of applying more fertilizer. In this study we found the average hired labor price was about US$1.40 a day, and we assumed it takes an extra day to apply 61 additional kilograms of fertilizer per hectare. We found fertilizer to be priced at $0.71 per kg on average in our data. As such the total cost of this increased production would be $0.71*61 kg + $1.40 = $45 per ha * 385,000 rented ha for a total cost of $17.33 million. This suggests that renting land generated a current year benefit/cost ratio of 1.36 (benefits of $23.5 million/costs of...
The cost of renting land is considered to be a transfer from tenants to landlords, and the net benefits need to be weighed against the potential longer-term soil fertility decline from renting land, as evidenced by tenants being 7 percentage points less likely to apply green compost to rented-in plots than their landlords on average (Table 5). More detailed estimates of the longer-term costs of soil fertility declines under reduced soil fertility investments is beyond the scope of our analysis but certainly merits further study.

9. Conclusions

This study used a unique data set on matched tenant-landlord pairs from four districts in Malawi to compare input use and soil fertility investment decisions among tenants’ rented-in plots, tenants’ owner-operated plots, and the owner-operated plots of their landlords. To our knowledge, this is the first study to do so, giving us greater insights into how tenant and landlord production decisions are affected by rental market participation. The latter has been particularly weak in the empirical literature because of low rates of observation of landlords relative to tenants in survey data (Chamberlin and Ricker-Gilbert 2016; Deininger, Xia, and Holden 2017). We also investigated the factors that affect landlords’ decisions on which plots to rent out and which to cultivate herself at the start of the season. Since we have multiple plots per household and per rental pair, we use rental-pair and household FEIs to identify our results.

Our findings were consistent with earlier studies from Malawi, which indicated that tenants were generally wealthier than landlords on all dimensions besides prerental landholding; that is, assets, savings and education (Chamberlin and Ricker-Gilbert 2016; Ricker-Gilbert et al. 2019). Tenants in our sample were more likely to apply annual inputs like herbicide and inorganic fertilizer on their owner-operated plots compared with landlords. However, tenants were less likely to make soil fertility–enhancing investments such as applying green compost on their rented-in plots, compared with their landlords. In addition, tenants were less likely to apply organic manure, green compost, or minimum tillage on rented-in plots than they were on their owner-cultivated plots. Previous studies have found that tenants were more likely to make longer-term soil conservation/enhancement investments on owner-operated plots relative to rented-in plots (Gavian and Fafchamps 1996; Jacoby and Mansuri 2008; Lovo 2016; Muraoka, Jin, and Jayne 2018). Our results add to this literature by comparing a variety of soil fertility investments on tenants’ rented-in plots, their owner-operated plots, and their landlord’s owner-operated plots, using our matched sample. Furthermore, it seemed that the major factor affecting whether a landlord rented out a plot was the presence of readily observable endowments, as landlords were 49–51 percentage points less likely to rent out a plot with fruit trees on it.

Our overarching interest has been to generate policy-relevant insights about the effects of rental market participation on soil fertility investments through a more comprehensive assessment of landlord and tenant sides of the market than has been done to date. Our comparison of rented and unrented plots operated by tenants is consistent with earlier studies suggesting that soil fertility investments are lower in rented-in plots compared with tenant’s owner-operated plots. However, our comparison of rented out versus landlord operated plots—which has largely been unexamined in prior literature—offers tentative evidence of differences in soil fertility management, primarily via lower use of green compost but not through the other investments evaluated. More empirical work, covering larger geographical diversity, will help clarify the net soil fertility management implications of expanding farmland rental markets in the longer term. The short-term gains from higher expected fertilizer investments by wealthier tenant farmers appear to be important, but figuring out how to incentivize greater fertility investments in farming systems that are increasingly defined by rental transactions is essential for sustainable productivity increases in the region.
Acknowledgments

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References


