RESEARCH ARTICLE

Land Ownership Security, Farm Investment, and Investment Risk in Indian Agriculture: Evidence from Nationally Representative Survey

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Abstract
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This paper examines the impact of land ownership on farm-level investment and investment risk (variability) in Indian agriculture using nationally representative survey data. The study adopts a flexible moment approach with White’s heteroscedasticity consistent standard error method. Results reveal that land ownership intensity significantly augments on-farm investment and reduces its variability. Other variables like family head’s education, access to irrigation, technical advice, credit and nonfarm income have significant inducement effects on-farm investment. We also observe that farmers’ education, age, irrigation, technical advice, MSP awareness and commercialization reduce farm investment risk. Findings have important policy implications for Indian land tenancy

Keywords: Farm investment; land ownership; Indian agriculture; NSSO data; investment risk

1. Introduction
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Agriculture is the primary source of livelihood in most developing countries. Enhancing agricultural production and productivity serves the vital policy objective of augmenting farmers’ income and alleviating poverty (Koirala et al., 2016; Christiaensen et al., 2011). Large-scale investment for infrastructural developments in this sector comes primarily from government initiatives (public sector investment) in irrigation, dams, power, storage, transport, marketing, research, and development activities. However, farmers and cooperatives (private sector) put their efforts and resources into creating farm assets and land conservation through land reclamation, bunding, wells, orchards and plantations, irrigation resources, agricultural implements, and farm machinery (Akber, Paltasingh, and Mishra, 2022; Kumar et al., 2020). The resources put for all these asset creations and conservation form private investments at the farm level by the farmers, and they are complemented and supplemented by public investment in farmland (Akber and Paltasingh, 2019, 2024; Gulati and Bathla, 2001).

Public and private investments are required for India’s agricultural sector’s productivity, sustainability, and growth (Akber and Paltasingh, 2021, 2024; Bathla and Kumari, 2017). However, when public investment has been secularly declining since the 1980s, the private investment at the farmers’ level is not up to the expected level (Gulati and Bathla, 2001; Chand & Kumar, 2004; Akber et al., 2022), rendering growth stagnation in Indian agriculture (Nadkarni 2018; Sainath, 2018; Akber and Paltasingh, 2021). So, the revival of that increasing trend in both types of investment holds the key to reversing the gloomy picture in Indian agriculture. However,
maintaining the right tradeoff between two types of public expenditure, i.e., public investment and farm input subsidies to boost on-farm private investment, is a matter of public policy (Akber et al., 2022). Knowing what stimulates farmers to put adequate efforts and investments into farmland at their level, except public farm investment, could be critical from a public policy perspective. Moreover, understanding the factors explaining the farm-level investment and its variability may provide the necessary inputs for better policymaking and implementation of designed policies.

The literature argues that among many factors (such as credit accessibility, farm size, farm experience, and farmers’ education), land ownership/tenure security is a vital institutional factor affecting farmers’ decisions to invest in farmland. Economic theory predicts that a lack of tenure security regarding land ownership or legal rights supporting users’ claims or tenancy rights may restrict and/or discourage tenants from investing optimally in farmland (Banerjee and Ghatak, 2004; Koirala et al., 2016; Paltasingh, 2018). It is because the tenants share a portion of their produce with the landlord and live under constant threat of being evicted from cultivating the land (Ali et al., 2012; Banerjee and Ghatak, 2004; Koirala et al., 2016; Paltasingh et al., 2022).

Lack of investment in farmland leads to a decline in farm productivity and efficiency—popularly known as the “Marshallian inefficiency hypothesis,” termed by Alfred Marshall in the early 1920s. The hypothesis has attracted much attention among researchers, and we have a vast empirical literature examining this hypothesis. However, researchers have no consensus on the empirical validation of the above hypothesis (Besley and Ghatak, 2010). Nevertheless, findings from some studies supported the arguments (Feder and Onchan, 1987; Shaban, 1987; Place and Hazell, 1993; Basley, 1995; Banerjee and Ghatak, 2004; Li et al., 1998; Shively, 1997; Jacoby and Mansuri, 2009; Abdulai et al., 2011; Ali et al., 2012; Bambio and Agha, 2018; Paltasingh, 2018; Paltasingh et al., 2022). These studies investigate the impact of land ownership security on farm-level investment, productivity, efficiency, and technology adoption. However, some other studies (Gavian and Ehui, 1999; Kassie and Holden, 2007; Pender and Fafchamps, 2006) contradict the Marshallian inefficiency hypothesis. Further, Migot-Adholla et al. (1994) and Pinkney and Kimuyu (1994) found land rights had a low impact on investments in soil improvement and the planting of tree crops. Lastly, Place and Hazell (1993) reveal, with few exceptions, that land rights were not significant determinants for investing in land.

In Indian agriculture, many micro-level studies have examined this hypothesis, but the debate remains inconclusive.2 This study aims at investigating the same hypothesis using nationally representative survey data on India’s agriculture sector. It also contributes to the literature by examining the role of land ownership in the variability of farm investment from its mean investment level, which is grossly ignored in Indian agriculture and across the globe. Hence, the main objective of this study is to examine the impact of land ownership and other socioeconomic factors on farm-level investment and variability in investment. The study used cross-sectional data compiled from the Situation Assessment Survey of Agricultural Households 2019–2021 of the 77th round of the National Sample Survey Office (NSSO).

The core analysis revealed that land ownership status induces a higher level of farm investment and reduces the variability of investment. Among other determinants, farm size, farmer’s education, age, access to irrigation, credit, technical assistance, awareness about the minimum support price (MSP), nonfarm income, etc., significantly induce on-farm investment. But, along with landownership status, farm size, farming experience, education, access to technical assistance, irrigation, knowledge about MSP, etc., reduce investment variability. In contrast, credit access, nonfarm income, and household size augment investment variability.

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1 The uncertainty of the users’ claims to the land and the crops they harvest may lessen expected future returns to investment (Gavian and Ehui, 1999).

2 Some studies such as Banerjee and Ghatak (2004), Paltasingh (2018), Shaban (1987), Sharma et al. (1995), Tripathy (1985), etc., argued that land ownership/secure tenancy augments farm productivity by crowding in more investment and efforts. However, very few studies such as Rao (1971), Chattopadhyay & Sarkar (1997), Chattopadhyay & Sengupta (2001) found contrasting results. Some studies like Vyas (1970) and Chattopadhyay (1979) could not find any significant conclusive empirical evidence.
This study contributes to the literature in several ways. First, it is the first study in the case of Indian agriculture covering a large sample, looking into the determinants of investment and investment risk (variability). So, it is unique as it gives insights not only into factors affecting mean investment level but also its variability. Second, the core results are obtained after controlling for major econometric problems like endogeneity and heterogeneity, which are relevant for effective policymaking by planners and policymakers for farm-level capital accumulation. This is vital for alleviating growth stagnation in Indian agriculture.

2. Reforms in land tenure system in India: a historical anecdote

Land reforms in India’s rural economy have a long history, even before the country’s independence. However, on the eve of India’s independence in 1947, land reforms, especially the redistribution of land from the haves to the have-nots, became a top priority on the policy agenda (Ghatak and Roy, 2007). Since independence, a significant body of land reforms has been enacted into law. The Constitution of India in 1949 granted the states the power to adopt and implement land reforms. This resulted in the execution of the reforms varying significantly across the states and over time. Land reform acts have been classified into four categories (Mearns, 1999; Basley and Burges, 2000). The first category is tenancy reform for regulating and stipulating the contract terms, namely shares in share tenancy contracts and attempts to abolish tenancy and transfer ownership to tenants. The second category abolished the intermediaries that prevailed before independence. The main goal of these acts was to curtail the powers of exploitative landlords. The third category imposed ceilings on the amount of land a person could own. The fourth category of land reform redistributed surplus land from the haves to the have-nots and consolidated landholdings. This reform act ensured that small landowners who owned small portions of land at different places could consolidate them into a single holding to boost productivity in the sector. These land reform measures undoubtedly eliminated some feudalistic characteristics in agrarian relations. Still, the country’s agricultural structure has not become more rational and equitable (Chattopadhyay, 1973).

Every Indian state passed land reform laws during the 1960s and 1970s to ensure greater tenure security. However, with some exceptions, like in Kerala and West Bengal, the results were not satisfactory (Ghatak and Roy, 2007). Due to weaknesses in the law and loopholes, the gap between the laws’ objectives and implementation was significant (Hanstad et al., 2008). For example, under the landholding ceiling law, 1.7% of the land was considered surplus, but only 1% was redistributed (Appu, 1996). The West Bengal Land Reforms Act of 1971 imposed the limit of 5 hectares of irrigated land as the ceiling for a family of five. In the case of unirrigated land, the ceiling was 7 hectares per family. Under this act, landowners were required to produce detailed records of their possessions, family size, and the surplus land they were supposed to surrender to the government. However, problems in implementing and executing this act soon became apparent (Bardhan and Mookherjee, 2010). Similarly, the institution of tenancy was abolished during the 1970s, and the states of Assam, Gujarat, Kerala, Tamil Nadu, and West Bengal implemented the policy effectively to give tenants more bargaining power. However, the self-interest of landlords forced the state governments to delay the legislation (Thimmaiah and Aziz, 1984).

Nonetheless, this policy helped the tenants and the actual cultivators to some extent in some places. Fearing the loss of their land, owners engaged in nonagricultural activities, leaving large areas uncultivated (Thimmaiah, 2001). Similarly, the policy of putting ceilings on land holdings was not uniform in all the states. In states like Andhra Pradesh, Bihar, Gujarat, Haryana, Jammu & Kashmir, Himachal Pradesh, Karnataka, Kerala, Rajasthan, and Tamil Nadu, the policy applies to

3By arguing that the agricultural structure has not become more rational and equitable, we essentially mean an insecure tenancy structure and a highly skewed distribution of landholding in India. Though these reforms aimed at reducing these loopholes but were not successful to a great extent.
both the owner and the tenant. However, only owners were subject to it in Assam, Maharashtra, Odisha, Punjab, Uttar Pradesh, and West Bengal. Ceiling limits also varied across the states.

The defects and loopholes in overall land reforms resulted in the land reform legislation harming India’s agricultural productivity (Ghatak and Roy, 2007). But as Srinivasan (2007) pointed out, due tenancy reform – Operation Barga of West Bengal – was instrumental in protecting tenants’ legitimate and legal rights and modestly augmented tenants’ productivity. Furthermore, Besley and Burgess (2000) showed that adopting land reforms (especially for tenancy protection legislation) in Indian states between 1958 and 1992 led to considerable decreases in rural poverty rates.

Looking at the landholding distribution with respect to ownership status, the latest round of NSSO data shows that 13% of total cultivable land is under tenancy while 85.6% is under ownership holdings. The trend of landownership from various rounds of NSSO data shows a declining share of ownership in operational landholding in India over the past two decades. Figure 1 shows that the percentage share of owned land in total operational holdings declined to 85.6% in 2018–19 from 92.7% during 2002–03. However, the share of leased-in land registered a two-fold increase (6–13%) during the given periods. So, this increasing trend of tenancy underscores the importance of studying its influence on farm-level investment and its variability, along with other farm and farmer-specific characteristics.

3. Theoretical framework

For examining the impact of land tenure security on farm-level investment and its variability, we develop a simple theoretical model based on the flexible moment approach within the expected profit maximization framework of Antle (1983, 87). The intention is to introduce a simple model necessary for useful data analysis. In the model, the farmer’s investment function typically is assumed to be a function of plot/farm size \((x)\), the farmer’s specific characteristics \((z)\), and the tenure security \((s)\) as:

\[
I = f(x, s, z),
\]

where “\(I\)” represents the farm-level investment, \(x\) is plot or farm size, \(s\) is tenure security/landownership status, and \(z\) is the vector of farm and farmers’ specific characteristics like soil quality, land position, irrigation accessibility, farming experience, education, farm income, household size, credit access, etc. In this setting, the source of investment risk (variability in investment) is assumed to be determined mainly by land ownership status, which is implicitly

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**Figure 1.** Trends in operational landholding distribution in Indian agriculture. Source: NSSO’s Situation Assessment Survey of Agricultural Households – 2003–04, 2013–14 and 2019–20.
Having defined the proportion of owned land to the total operational landholding. In that case, it implicitly paves the way for the farmer’s adequate/optimal investment efforts. Following Paltasingh (2018) and Paltasingh et al. (2022), we accommodate all the farmers while defining this landownership status \(s_i\) for the \(i\)th farmer as:

\[
s_i = 1 - \gamma_i,
\]

where \(\gamma_i = \frac{l}{O_i}\) measures the proportion of leased-in land \((l)\) in aggregate operational landholding \((O_i)\) of that \(i\)th farmer, indicating tenancy intensity. Thus, the value of \(s_i\) varies within the range of 0 and 1. A value of 1 indicates that the concerned farmer is a pure owner-operator whose ownership status is highly secured. These farmers have complete ownership status in their operational holdings. A value of 0 indicates the reverse, as the entire operational landholding consists of leased-in land. The distribution of \(s_i\) expressed as \(s \sim \chi(s|\gamma)\) and is assumed to be a major determinant of farm-level investment and investment variability. Following Antle (1983, 1987) and Zhang and Antle (2016), the variability in farm-level investment is captured by introducing the variance of investment through the moment-based approach as follows:

\[
E[(f(x, s, z) - f_1(x, s, z))^2] = f_k(x, s, z, \beta_k) \forall k \geq 2,
\]

where \(f_1(.) = E(f(x, s, z))\) represents the mean of the investment function. Given equation (3), the first moment of the investment function can be defined as:

\[
E[f(x, s, z)] = f_1(x, s, z, \beta_1) = \mu_1.
\]

Similarly, the second moment (variance) of the investment function is defined as:

\[
E[(f(x, s, z) - E(f(x, s, z)))^2] = f_2(x, s, z, \beta_2) = \mu_2.
\]

Following Zhang and Antle’s (2016) specifications, in equations (4) to (5), we can express \(E[U(\pi)]\) as a function of the first two moments of the distribution of net returns from investment:

\[
E[U(\pi)] = U[f_1(x, s, z, \beta_1), f_2(x, s, z, \beta_2)] = U(\mu_1, \mu_2),
\]

where \(\pi\) is the net returns from investment and \(E\) is the expectation operator. We assume farmers are risk-averse and maximize the expected utility of net returns from investment by choosing \(I\) and \(x\) in the following manner:

\[
\max_{I,x} E[U(\pi)] = \max_{I,x} \int U(\pi) \chi(s|\gamma) dT = U(\mu_1, \mu_2)
\]

Having defined \(U^j = \frac{\partial U(\mu_1, \mu_2)}{\partial \pi_j}; j \in \mathbb{N}\), and using a second-order Taylor’s approximation of the expected utility function yields the optimality condition as:

\[
\frac{\partial \mu_1}{\partial x} + \frac{1}{2} \frac{U''(\pi)}{U'(\pi)} \frac{\partial \mu_2}{\partial x} = 0.
\]

This optimality condition in equation (8), in the form of elasticity, can be written as:

\[
\mu_1^* = Rv_2\mu_2^*
\]

Where \(\mu_j^* = \frac{\partial \mu_j}{\partial x}\) and \(v_2 = \frac{\mu_2}{\mu_1}\) is variance (second central moment of \(\pi\)). \(R = \frac{1}{2} \left[ \frac{U''(\pi)}{U'(\pi)} \right] \) is the Arrow-Pratt relative risk aversion coefficient, which represents the risk attitude or the marginal risk premium, and measures the farmer’s willingness to substitute a change in the expected net returns for a change in investment risk or investment variability.
The certainty equivalent (CE) of the expected utility function can be written as:

$$E[U(\pi)] = U(CE) = U(\mu_1 - R). \quad (10)$$

So, $R$ is the risk premium the farmer is willing to pay to avoid the variability in investment returns. This can be considered as the cost of private risk bearing or avoiding investment risk. Since the utility is positively related to returns from investment, the optimization problem in equation (7) is equivalent to maximizing the CE, and the first-order optimality condition is expressed as:

$$\frac{\partial \mu_1}{\partial x} = \frac{\partial R}{\partial x} \quad (11)$$

In combination with equation (8), the differentiation of marginal risk premium $R$ with respect to $x$ results in:

$$\frac{\partial R}{\partial x} = R \cdot \frac{\partial \mu_2}{\partial x} \quad (12)$$

Equation (12) shows the effect of farm size on investment variability and measures the total impact on risk premiums. If plot size increases the investment variability, it has a positive marginal risk premium and vice-versa (Chavas, 2004). The above can be interpreted as the marginal risk effect of farm size, equal to the marginal mean net returns from investment. A farmer chooses the optimal farm size when equality occurs between marginal net returns from farm size and its marginal risk effect. The solution to this optimality condition gives us:

$$\begin{cases} 
  x^* = x(z, s) \\
  I^* = I(x^*, z, s)
\end{cases} \quad (13)$$

The first part of equation (13) shows that the demand for bringing more land under operational holding by a farmer is a function of $z$, the vector of farmer-specific characteristics and site-specific factors, and the landownership status $(s)$. The second part shows that the decision to put the optimal investment level depends on $x$, $z$, and landownership status(s).

### 4. Data and empirical framework

This study used the cross-sectional data compiled from India’s 77th round (January 2019–December 2019) of NSSO’s Situation Assessment Survey of Agricultural Households (SASAH, 2019–21). The current round of NSSO has a sample of 45,714 farming households at the national level. Farm investment data was compiled by adding the expenses incurred on purchasing significant repair/improvement in the land, building for farm business, and machinery. The explanatory variables include land ownership status, farm size, farmers’ age, education, total credit (loan amount in cash or kind), access to irrigation, technical advice/assistance, awareness of minimum support price (MSP), access to crop insurance, rate of market orientation, and nonfarm income. The rate of market orientation is defined as the share of total crop output sold (in the local market, mandis, or through any other agencies) to the total value of the crop produced. Table 1 presents the definitions and

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4 Though the survey was conducted during January 2019 to December 2019, the report and the unit-level data were released in 2021.

5 Farm machineries include sickle, chaff cutter, axe, spade, chopper, plow, harrow, power tiller, tractor, threshers, cane, crusher, oil crusher, pump, and other water-lifting equipment.

6 Defined as $(s_i = 1 - \gamma_i)$, where $\gamma_i$ the ratio of land leased-in to the total operational land, indicating the ownership intensity in total operational landholding. The values vary between 0 and 1. While 0 indicates pure tenants, the value of 1 indicates owner-operator. The specific terms of lease on which data were collected in the survey are: (i) for fixed money, (ii) for fixed produce, (iii) for share of produce, (iv) from relatives under no specified terms, and (v) under other.
descriptive statistics of the variables used in this study. We transformed the discrete variables into natural logarithms to avoid the possible skewed distribution of those variables.\footnote{The logarithmic transformation is widely used method to avoid skewed distribution of the data. After having logarithm transformation, we actually get significantly small amount of positive or negative skewness. Using the formula of skewness $S_k = 3(Mean - Median)$, we get the value of $S_k$ as 0.11 for farm investment. So, there is a slight positive skewness in the variable. Similarly, very small positive or negative skewness or no skewness results are also obtained for other variables. Hence, the logarithmic transformation significantly controls the skewness in the data.}

### 4.1. Empirical strategy

We adopt an empirical strategy based on our theoretical framework to understand the impact of land ownership and other socioeconomic and farm-specific variables on farm-level investment and investment variability. Since the theoretical model is based on the flexible moment approach developed by Antle (1983), where landownership is incorporated directly in the first and second moments, the empirical strategy is also based on this method. Thus, the impact of land ownership

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
<th>Mean</th>
<th>Std. Dev.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Dependent variable</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>FI</td>
<td>Private on-farm investment (annual expenditures on farm assets purchased by the households, \textit{in Rupees/ha})</td>
<td>4628.42</td>
<td>53353</td>
</tr>
<tr>
<td><strong>Independent variables</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>OL</td>
<td>Land ownership status (index values vary between 0 and 1)\textsuperscript{a}</td>
<td>0.252</td>
<td>0.928</td>
</tr>
<tr>
<td>FS</td>
<td>Farm size (ha)</td>
<td>1.543</td>
<td>0.498</td>
</tr>
<tr>
<td>CRDT</td>
<td>Total credit received by farmers yearly (in cash or kind, \textit{in Rupees})</td>
<td>17395</td>
<td>27234</td>
</tr>
<tr>
<td>AGE</td>
<td>Age of the farmers (years)</td>
<td>28.59</td>
<td>19.29</td>
</tr>
<tr>
<td>EDN</td>
<td>Education of the farmer(years of schooling)</td>
<td>4.734</td>
<td>5.980</td>
</tr>
<tr>
<td>IR</td>
<td>Access to irrigation (= 1 if land is irrigated; 0 = otherwise)</td>
<td>0.608</td>
<td>0.309</td>
</tr>
<tr>
<td>TA</td>
<td>Access to technical assistance from agricultural officers (= 1 if yes; 0 = otherwise)</td>
<td>0.796</td>
<td>0.292</td>
</tr>
<tr>
<td>MSP</td>
<td>Awareness about \textit{minimum support price} (MSP) (= 1 if yes; 0 = otherwise)</td>
<td>0.828</td>
<td>0.377</td>
</tr>
<tr>
<td>NFI</td>
<td>Nonfarm income (annual total earnings from nonfarm activities, \textit{in Rupees})</td>
<td>346893</td>
<td>37771</td>
</tr>
<tr>
<td>MO</td>
<td>Market orientation (percentage of marketed value of crop output to the total value of crop production)\textsuperscript{b}</td>
<td>78.1</td>
<td>3.30</td>
</tr>
<tr>
<td>HS</td>
<td>Household size (number)</td>
<td>5.359</td>
<td>2.692</td>
</tr>
<tr>
<td>CI</td>
<td>Crop insurance (= 1 if access to crop insurance; 0 = otherwise)</td>
<td>0.628</td>
<td>0.792</td>
</tr>
<tr>
<td>AO</td>
<td>Annual value of output produced by/from animals, like egg, milk, meat, fish, honey, etc. (\textit{in Rupees})</td>
<td>6131</td>
<td>16337</td>
</tr>
<tr>
<td>NCPH</td>
<td>Non-cultivating peasant households (= 1 if yes; 0 = otherwise)</td>
<td>0.012</td>
<td>0.172</td>
</tr>
<tr>
<td>N</td>
<td>Total observations of farming households</td>
<td>45714</td>
<td></td>
</tr>
</tbody>
</table>

Source: NSSO's 7\textsuperscript{th} round – Situation Assessment Survey of Agricultural Households (SASAH), 2019.

\textsuperscript{a} Defined as the share of land leased-in. The values vary between 0 and 1. Here, 0 indicates pure tenants, and 1 indicates owner-operator. The specific terms of lease on which data were collected in the survey are: (i) for fixed money, (ii) for fixed produce, (iii) for share of produce, (iv) from relatives under no specified terms, and (v) under other terms.

\textsuperscript{b} The share is multiplied by 100 to give a percentage, and we add one to zero values (no market orientation). In logarithmic transformation, this value comes as zero. In normalization process, we add one to zero value and then take log transformation.
status and other socioeconomic variables on the mean investment function and investment variability is examined through the first and second moments.\(^8\)

The following model (Equation 14) estimates the mean investment function as follows:

\[
I_i = f(x_i, z_i, s_i, \theta_i) + u_i
\]  
(14)

where \(x_i\) refers to farm size, \(z_i\) refers to vectors of farm and farmer-specific characteristics, \(s_i\) refers to land ownership status and \(u_i\) is the error term. Finally, \(\theta_i\) is a vector of parameters to be estimated. To capture the variability of farm-level investment, the estimated errors from the above equation are obtained, and the squared values are regressed on the same explanatory variables to estimate the second moment of the farm investment as:

\[
\hat{u}_i^2 = g(x_i, z_i, s_i, \vartheta_i) + \epsilon_i
\]  
(15)

The estimation of the vector of parameters \(\vartheta_i\) in this model will show the effect of explanatory variables on the variability of farm investment. A risk-reducing factor carries a negative sign, and a risk-enhancing factor will have a positive coefficient.

### 4.2. Endogeneity issue

Many studies assume that land ownership status is exogenous because the land rights determine it. However, some recent studies (Ali et al., 2012; Brasselle et al., 2002; Paltasingh et al., 2022; Place and Otsuka, 2002) argue that tenancy arrangements and productivity-enhancing investments are jointly determined. Thus, it leads to the problem of endogeneity of land ownership status/tenure variable in equation (14). Hence, we need to test for the potential endogeneity of land ownership status. Following Wooldridge (2015), we use the “control function approach” where, in the first step, we model landownership status \((s_i)\) as a function of vectors of various explanatory factors \((z_i\) and \(x_i)\) and the instrument variable \((IV_i)\) as given in equation (16):

\[
s_i = h(IV_i, z_i, x_i, \varphi_i) + \omega_i
\]  
(16)

where \(\varphi_i\) is the parameter vector to be estimated and \(\omega_i\) is an error term. In the second step, the test involves writing the error term \((u_i)\) in equation (14) as conditional on \(\omega_i\), specifically as:

\[
u_i = \omega_i \phi_i + v_i
\]  
(17)

Substituting \(u_i\) in equation (14) yields the conditional model as:

\[
I_i = f(s_i, z_i, x_i, \theta_i) + \omega_i \phi_i + v_i
\]  
(18)

The test of endogeneity of \(s_i\) is performed by testing the null hypothesis, \(H_0: \phi_i = 0\). Failing to reject the null hypothesis implies that the land tenancy variable is exogenous. The test involves estimating equation (16) with the control function approach for endogeneity (Wooldridge, 2015), where the instrument is modeled as one of the determinants of land tenure security, along with other explanatory variables. Then, we estimated the equation (18) in the second stage using the residuals in the first stage.\(^9\) If the coefficient \(\phi\) is significant, then the exogeneity of \(s_i\) is rejected. The above method involves identifying an instrument for landownership status. The variable excluded from the second-stage regression model is a dummy variable indicating the household is a “non-cultivating peasant household” (NCPH). An NCPH earns some portion of its income or

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\(^8\)A feasible generalized least square method developed by Just and Pope (1978, 1979) is another close variant of flexible moments approach. It can also be used for the same purpose of finding the source of variability in investment.

\(^9\)In the second stage, equation (18), i.e., mean investment function is estimated with bootstrapped standard error, as White’s robust standard errors do not correct for the use of generated regressors on mean investment equation in final stage. But, in the final estimation, we used heteroscedasticity consistent White’s robust standard errors.
total income from nonfarm activities but not from cultivation. We argue that this is a good instrument because NCPHs are the households that own land but never cultivate it (Vijay, 2012). As Vijay (2012) and Bhue and Vijay (2012) pointed out, a new class of landowners emerging in Indian agriculture own the land but do not cultivate it. Their primary source of income is nonfarming activities. One can argue that these are absentee landlords who mainly lease out their farm land, thus increasing the incidence of tenancy in Indian agriculture. Therefore, the NCPH, being a relevant instrument, also satisfies the required exclusion restriction condition because being an NCPH does not affect investment directly.

Tables 1A and 2A in the appendix present the first and second-stage results of the endogeneity test. The first-stage regression model (Eq. 16) shows the effects of household and farm-level characteristics and the instrument (NCPH) on land ownership status. We found that the dummy variable, NCPH, is negative and significantly influences landownership status, indicating that these households are more likely to lease out the land. In other words, land owned by families but not cultivated by them is more likely to be leased out, increasing tenancy in Indian agriculture. Using the F-statistic to evaluate the weakness of this instrument, we observe that the instrument is not weak as the F-statistic is sufficiently large. The estimated results of the second-stage model (Equation 18) presented in Appendix Table 2A indicate that the residual (\(u\)) derived from the first stage is insignificant.

5. Results and discussion

A commonly used method of testing the Marshallian hypothesis is to compare on-farm investment under land ownership. This comparison provides a pretest to that hypothesis before the multivariate regression analysis. Here, we present the mean difference test (\(t\)-test) of farm-level investment and other inputs between owner-cultivators and tenants and, within the tenancy structure, between pure and partial tenants. Table 2 shows the mean difference between tenants and owner-cultivators. The farm-level mean investment under ownership cultivation is significantly higher than that of sharecroppers/tenants. This significant mean difference in farm-level investment between the two groups provides direct evidence that ownership security is positively associated with on-farm investment. Similarly, there are significant mean differences in the input uses and other characteristics. The significant mean difference is also observed among the inputs between farm size, credit access, and degree of market orientation of tenants and owner-cultivators. Regarding farm size, tenants’ mean operational landholding is substantially greater than owner-operators. But in all other cases, the owner-cultivators are better off than tenants.

The owner-operators have significantly more access to credit than tenants have. This result aligns with Feder and Onchan (1987) and Ali et al. (2012). They argued that access to institutional credit depends significantly on the capacity to put up collateral associated directly with land ownership. Likewise, the rate of market orientation is higher for owner-operators than tenants. Thus, land tenure security induces more farm investment and efficiency. Market participation

\[10^{th}\text{Ali et al. (2012) used two instruments in their field survey data. These include whether cultivator’s residence is located in the same village, and distance of plots from cultivators’ residence. However, in the NSSO’s survey data, this information on location and distance of plots from the residence is not available. Hence, we take this instrument for tenancy arrangement. They argued that distant plot raises the cost of self-cultivation which leads to renting out of plots. Again, greater distance between the plot and landowner’s residence also raises the cost of monitoring if leased out, resulting in potential biased estimates (Jacoby and Mansuri, 2009).}

\[11^{th}\text{NCPH are households who earn income from nonfarm activities like pension, remittances, nonfarm self-employment, small traders, input dealers, artisans. They exert pressure on land market as they own the land and lease out to tenants. But, NCPH do not cultivate the land. However, they benefit from investments on land by tenants.}

\[12^{th}\text{Bhue and Vijay (2012) found that around 4% of total cultivable land in Indian agriculture is under this NCPH in 2013, and it is rising over time.}

\[13^{th}\text{The rule of thumb is that F-statistic needs to be more than 10 for the instrument to be considered not weak (Hansen, 2022; p. 404). The value of F-statistic in this case correspond to the square of the value for the coefficient of the instrument in the 1st stage regression (Table 1A), i.e., (t-value of instrument)^2 = (-8.618)^2 = 74.27.}
positively determines gross returns or vice-versa, thus stimulating farm investment. Landownership status encourages farm-level investment through various direct or indirect channels. There is no significant difference between owner-operators and tenants in farmers’ education and farming experience (age as a proxy).

Table 3 presents the same comparison within the tenancy structure between pure tenants and partial tenants. Here, we consider partial tenants more secure than pure tenants because of their ownership rights in total operational landholding. The difference in on-farm investment between the two groups is about 5922.37 Indian rupees. Among other input variables, there is a negative and significant mean difference between pure and partial tenants in farm size, access to credit, education, and rate of market orientation. The analysis indicates that partial tenants have a larger average farm size, greater access to credit, and relatively greater market participation than pure tenants. Only in nonfarm income do pure tenants have a higher average nonfarm income than partial tenants. However, the above simple comparison of means analysis suffers from a limitation. It does not include causal relationships between land tenure security and farm investment, and endogeneity issues have not been addressed. Further, the above analysis sheds no light on the effects of land tenure security on the variability in on-farm investment.

Table 2. Mean difference in key characteristics between owners and tenants

<table>
<thead>
<tr>
<th>Variable</th>
<th>Owner</th>
<th>Tenant</th>
<th>Difference</th>
<th>t-Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>Farm investment</td>
<td>13905.15</td>
<td>13108.52</td>
<td>796.63***</td>
<td>2.852</td>
</tr>
<tr>
<td>Farm size</td>
<td>0.543</td>
<td>1.158</td>
<td>-0.615***</td>
<td>49.152</td>
</tr>
<tr>
<td>Credit accessed</td>
<td>34632.7</td>
<td>23748.8</td>
<td>10910.7***</td>
<td>28.564</td>
</tr>
<tr>
<td>Farmer’s age</td>
<td>25.754</td>
<td>25.731</td>
<td>0.238</td>
<td>0.145</td>
</tr>
<tr>
<td>Farmer’s education</td>
<td>5.002</td>
<td>4.997</td>
<td>0.004</td>
<td>0.155</td>
</tr>
<tr>
<td>Nonfarm income</td>
<td>37769.69</td>
<td>16915.67</td>
<td>20854.02</td>
<td>1.046</td>
</tr>
<tr>
<td>Market orientation</td>
<td>0.8045</td>
<td>0.0909</td>
<td>0.7125***</td>
<td>-4.853</td>
</tr>
<tr>
<td>N (Total households)</td>
<td>39131</td>
<td>6583</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: *** P < 0.01, ** P < 0.05, * P < 0.10. Units of variables are given in Table 1.
Source: Authors’ estimation from NSSO’s 7th round – Situation Assessment Survey of Agricultural Households (SASAH), 2019.

Table 3. Mean difference in key characteristics between pure tenants and partial tenants

<table>
<thead>
<tr>
<th>Variable</th>
<th>Pure Tenant</th>
<th>Partial Tenant</th>
<th>Difference</th>
<th>t-Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>Farm investment</td>
<td>11004.3</td>
<td>16926.68</td>
<td>-5922.37***</td>
<td>-3.832</td>
</tr>
<tr>
<td>Farm size</td>
<td>1.006</td>
<td>1.677</td>
<td>-0.671***</td>
<td>-21.46</td>
</tr>
<tr>
<td>Credit accessed</td>
<td>12900.4</td>
<td>25644.36</td>
<td>-12740.36***</td>
<td>-11.247</td>
</tr>
<tr>
<td>Farmer’s age</td>
<td>25.746</td>
<td>25.129</td>
<td>0.617**</td>
<td>2.282</td>
</tr>
<tr>
<td>Farmer’s education</td>
<td>5.13</td>
<td>5.513</td>
<td>-0.382***</td>
<td>-7.897</td>
</tr>
<tr>
<td>Nonfarm income</td>
<td>45705.96</td>
<td>20671.88</td>
<td>25037.19*</td>
<td>1.794</td>
</tr>
<tr>
<td>Market orientation</td>
<td>0.0337</td>
<td>0.06831</td>
<td>-0.0346***</td>
<td>5.509</td>
</tr>
<tr>
<td>N (Total households)</td>
<td>1829</td>
<td>4754</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: *** P < 0.01, ** P < 0.05, * P < 0.10. Units of variables are given in Table 1.
Source: Authors’ estimation from NSSO’s 7th round – Situation Assessment Survey of Agricultural Households (SASAH), 2019–21.

positively determines gross returns or vice-versa, thus stimulating farm investment. Landownership status encourages farm-level investment through various direct or indirect channels. There is no significant difference between owner-operators and tenants in farmers’ education and farming experience (age as a proxy).

Table 3 presents the same comparison within the tenancy structure between pure tenants and partial tenants. Here, we consider partial tenants more secure than pure tenants because of their ownership rights in total operational landholding. The difference in on-farm investment between the two groups is about 5922.37 Indian rupees. Among other input variables, there is a negative and significant mean difference between pure and partial tenants in farm size, access to credit, education, and rate of market orientation. The analysis indicates that partial tenants have a larger average farm size, greater access to credit, and relatively greater market participation than pure tenants. Only in nonfarm income do pure tenants have a higher average nonfarm income than partial tenants. However, the above simple comparison of means analysis suffers from a limitation. It does not include causal relationships between land tenure security and farm investment, and endogeneity issues have not been addressed. Further, the above analysis sheds no light on the effects of land tenure security on the variability in on-farm investment.

141 US $ = 83.06 Indian rupees (₹) in January 2024.
5.1. Determinants of farm investment

After finding no clear evidence of the endogeneity problem of land ownership status, we separately estimated the mean investment function and its variability. Tables 4 and 5 depict the results of determinants of on-farm investment function by estimating equation (14) and the determinants of farm investment variability in the second stage (Equation 15) after obtaining the second central moment. The statistics on the model’s fit (R-squared value and F-statistic) indicate a good overall model fit. The estimation procedure adopts White’s heteroscedasticity consistent robust standard errors. This is an appropriate method in a large cross-sectional sample that controls the heteroscedasticity problem. Note that explanatory and dependent variables are all transformed into natural logarithmic forms except the dummy variables. Results reveal that land ownership has a positive and statistically significant impact on the mean on-farm investment.\(^{15}\) The coefficient of land ownership status is positive (0.248) and statistically significant, indicating that a 1% increase in land ownership (i.e., the intensity of land tenure security) increases on-farm investment by about 0.25%. Our finding confirms the “Marshallian inefficiency” hypothesis that tenure security induces more investment and supply of efforts from cultivators, which augments farm efficiency and productivity. Our result also is consistent with Cao, Bai, and Zhang (2020), Paltasingh (2018), Ma et al. (2018), Abdulai, Owusu, and Goetz (2011), Deininger and Ali (2008), and Besley (1995).

\(^{15}\)Though, we got that landownership is exogeneous from the test, but in the IV procedure in the control function approach test for endogeneity, the coefficient of landownership status is statistically significant and the size of coefficient has slightly smaller (0.230) than its coefficient (0.248) in the main result. This confirms the importance of land ownership status as a determinant of farm level investment.

### Table 4. Determinants of farminvestment in Indian agriculture

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>Std. Error</th>
<th>Prob.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Land ownership status (ln)</td>
<td>0.248***</td>
<td>0.034</td>
<td>0.000</td>
</tr>
<tr>
<td>Farm size (ln)</td>
<td>0.128***</td>
<td>0.013</td>
<td>0.000</td>
</tr>
<tr>
<td>Credit accessed (ln)</td>
<td>0.212***</td>
<td>0.015</td>
<td>0.000</td>
</tr>
<tr>
<td>Farmer’s Age (ln)</td>
<td>0.016**</td>
<td>0.007</td>
<td>0.023</td>
</tr>
<tr>
<td>Farmer’s education (ln)</td>
<td>0.170***</td>
<td>0.004</td>
<td>0.000</td>
</tr>
<tr>
<td>Irrigation access (dummy)</td>
<td>0.116**</td>
<td>0.051</td>
<td>0.023</td>
</tr>
<tr>
<td>Technical assistance (dummy)</td>
<td>0.425***</td>
<td>0.067</td>
<td>0.000</td>
</tr>
<tr>
<td>Awareness about MSP (dummy)</td>
<td>0.255***</td>
<td>0.067</td>
<td>0.000</td>
</tr>
<tr>
<td>Nonfarm income (ln)</td>
<td>0.106***</td>
<td>0.015</td>
<td>0.000</td>
</tr>
<tr>
<td>Household size (ln)</td>
<td>−0.114***</td>
<td>0.002</td>
<td>0.000</td>
</tr>
<tr>
<td>Market orientation (ln)</td>
<td>0.276***</td>
<td>0.065</td>
<td>0.000</td>
</tr>
<tr>
<td>Crop insurance (dummy)</td>
<td>0.386**</td>
<td>0.176</td>
<td>0.028</td>
</tr>
<tr>
<td>Animal output (ln)</td>
<td>0.231***</td>
<td>0.016</td>
<td>0.000</td>
</tr>
<tr>
<td>Constant</td>
<td>0.044</td>
<td>0.225</td>
<td>0.844</td>
</tr>
<tr>
<td>F-statistic</td>
<td>20.61</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Prob(F-statistic)</td>
<td>0.000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>R-squared</td>
<td>0.910</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: ***\(P < 0.01\), **\(P < 0.05\), *\(P < 0.10\). The standard errors are White’s heteroscedasticity consistent standard errors and ln refers to natural logarithm.

Source: Authors’ estimation from NSSO’s 77th round on Situation Assessment Survey of Agricultural Households (SASAH), 2019-21. Dependent variable = \(\ln\) (on-farm expenditures on farm assets).
Cao et al. (2020) empirically revealed the significant positive effects of secure farmland property rights on land attached and long-term investments in land in rural China. Paltasingh (2018) examined Besley’s hypothesis (tenure security effect) in the agriculture of eastern India and found significant inducement effects of land tenure security on farm-level investment in adopting modern rice technology. Some other studies examined the same hypothesis in the case of African countries like Ghana, Uganda, and Burkina Faso and found similar results (Bambio and Agha, 2018; Besley, 1995; Ngom et al., 2023; Place and Otsuka, 2002; Donkor et al., 2023). Our study confirms this hypothesis by using the pan-Indian data.

The coefficient on education is positive and significant at the 1% level of significance. Results suggest that an additional 1% increase in years of schooling of farm household head increases on-farm investment by about 0.17%. Education helps farmers make sound and effective decisions, leading to more on-farm investment. Additionally, formal education is critical for adopting modern technology, farm production, farm management, and managerial skills (Paltasingh and Goyari, 2018; Paltasingh, 2016). The coefficient of farm size is positive and statistically significant at the 1% significance level, suggesting that an additional hectare of land increases on-farm investment by about 0.128%. Our conclusion is consistent with Feder et al. (1992) and Abdulai et al. (2011).

The coefficient on credit amount is statistically significant at the 1% significance level. The estimated elasticity coefficient of 0.212 indicates that a 10% increase in credit increases on-farm investment by 2.12%. Our finding is consistent with Petrick (2004), who found that institutional credit to the agricultural sector positively and significantly impacted private-sector investment. The farmer’s age has a positive and significant effect on farm-level investment. An explanation is that

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>S.E.</th>
<th>Prob.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Land ownership status (ln)</td>
<td>−0.082***</td>
<td>0.026</td>
<td>0.000</td>
</tr>
<tr>
<td>Farm size(ln)</td>
<td>−0.018*</td>
<td>0.010</td>
<td>0.072</td>
</tr>
<tr>
<td>Credit accessed (ln)</td>
<td>0.061**</td>
<td>0.037</td>
<td>0.019</td>
</tr>
<tr>
<td>Farmer’s Age(ln)</td>
<td>−0.041</td>
<td>0.031</td>
<td>0.996</td>
</tr>
<tr>
<td>Farmer’s education (ln)</td>
<td>−0.012**</td>
<td>0.006</td>
<td>0.048</td>
</tr>
<tr>
<td>Irrigation access (dummy)</td>
<td>−0.140*</td>
<td>0.008</td>
<td>0.080</td>
</tr>
<tr>
<td>Technical assistance (dummy)</td>
<td>−0.027*</td>
<td>0.014</td>
<td>0.072</td>
</tr>
<tr>
<td>Awareness about MSP (dummy)</td>
<td>−0.082**</td>
<td>0.042</td>
<td>0.049</td>
</tr>
<tr>
<td>Nonfarm income(ln)</td>
<td>0.042**</td>
<td>0.020</td>
<td>0.036</td>
</tr>
<tr>
<td>Household size (ln)</td>
<td>0.012**</td>
<td>0.006</td>
<td>0.044</td>
</tr>
<tr>
<td>Market orientation (ln)</td>
<td>−0.020***</td>
<td>0.007</td>
<td>0.004</td>
</tr>
<tr>
<td>Crop insurance (dummy)</td>
<td>0.032</td>
<td>0.077</td>
<td>0.682</td>
</tr>
<tr>
<td>Animal output (ln)</td>
<td>−0.010</td>
<td>0.031</td>
<td>0.733</td>
</tr>
<tr>
<td>Constant</td>
<td>0.154</td>
<td>0.486</td>
<td>0.751</td>
</tr>
<tr>
<td>F-stat</td>
<td>4.412</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Prob(F-statistic)</td>
<td>0.000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>R-squared</td>
<td>0.280</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: ***P < 0.01, **P < 0.05, *P < 0.10. The standard errors are White’s heteroscedasticity consistent standard error; ln is the natural logarithm.

Source: Authors’ estimation from NSSO’s 77th round – “Situation Assessment Survey of Agricultural Households” (SASAH), 2019–21.

Dependent variable = ln (Squared residual).

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farmers tend to accumulate farming experience, knowledge, and wealth as they age. In addition, experienced and successful farmers know the farming situation in the area and are interested in expanding operations. In addition, Martey et al. (2012) argue that older household heads might have more contacts, allowing them to find trading partners or technology to lower production costs.

The coefficient of access to surface irrigation is positive and statistically significant at the 1% level of significance. Results suggest that access to irrigation increases private on-farm investment by 0.116%. Our finding is consistent with Pender and Kerr (1998), who found that access to irrigation had a significant effect on small holders’ investments in soil and water conservation methods in the semi-arid region of India. Awareness of MSP policy positively and significantly impacts the on-farm investment. Indeed, government support in farming in the form of guaranteed MSP for commodities signifies farm income stability, thus reducing riskiness in on-farm investment. Aditya et al. (2017) argue that MSP provides security for long-term investment. Nonfarm income has a positive and highly significant impact on on-farm investment. This positive impact is in line with Ma et al. (2018), who empirically found that the effect of off-farm work stimulates agricultural production by increasing investments in productivity-enhancing inputs. For instance, Zhang et al. (2020) found that off-farm employment increased the rate of chemical fertilizer intensity in plains areas of China.

The coefficients on the rate of market orientation and value of animal output (meats, milk, fish, honey, eggs) are positive and significant at the 1% level of significance. Results indicate that a 1% increase in the rate of market orientation and value of animal output increases on-farm investment by about 0.28% and 0.23%, respectively. The result of the positive and significant effects of the rate of market orientation is consistent with Strasberg et al. (1999), who found that agricultural household’s farm commercialization has a significant and positive impact on fertilizer use and productivity. On the other hand, as expected, household size negatively and significantly affects smallholders’ on-farm investment in India.16

5.2. Determinants of investment variability
Table 5 presents parameter estimates of factors affecting investment variability at the farm level of Indian farmers.17 A positive coefficient enhances on-farm investment variability, and a negative coefficient reduces on-farm investment variability. The negative coefficient of land ownership status (−0.08) is statistically significant at a 1% probability level. Results indicate that 1% rise in landownership status/security reduces the on-farm investment variability by 0.082%. The relative strength of the coefficient is small, but what is important is that secured land ownership rights reduce on-farm investment risks. Our finding is consistent with Besley (1995), Demsetz (2003), and Deininger and Jin (2006), who found that farmland property rights decreased transaction costs and investment risk. Among the other significant factors that reduce the farm investment variability are farm size (−0.018), farmer education (−0.012), technical advice (−0.027), irrigation dummy (−0.140), awareness about MSP (−0.082), and market orientation (−0.020). Thus, findings reveal that educated smallholders who received technical advice and assistance maintain a persistent level of investment, reducing the variability of on-farm investment more than their counterparts. Farmers who seek technical advice are usually progressive farmers seeking ways to improve the viability and profitability of their farming operations (Paltasingh and Goyari, 2018). Assured irrigation reduces production risk, stimulating on-farm investment and reducing its variability. Knowledge about the market and participation in the market also minimizes investment risks. The rate of market participation and awareness about crop prices encourages

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16Smallholders are mainly marginal and small farmers having landholding less than one hectare. In Indian agriculture, their percentage is around 87%.
17The squared of residuals obtained from the first stage is used as the estimates for the second central moment, giving us the on-farm investment variability.
farmers to invest in the farm and increase profits, thereby reducing the variability in on-farm investment. Among all these variability-reducing factors, access to irrigation has a relatively bigger risk-reducing effect.

The estimated coefficients of credits (0.061), nonfarm income (0.042), and household size (0.012) are positive and statistically significant at the 5% significance level. Results suggest that loans or credit amounts enhance the on-farm investment variability. This may be because loans need to be repaid in the future, thus requiring farmers to think hard about investing additional resources in on-farm investment. Any additional funds or income would go toward the repayment of loans first. The coefficient of household size confirms that an increase in the farmers’ family size enhances the variability in farm investment. The coefficient is positive at 0.012 and statistically significant, indicating its variability-enhancing effect. This can be attributed to increased consumption expenditures associated with increased household size. This causes variability in farm investment as increased consumption expenditure affects farm investment in both ways. First, it siphons scanty resources off for required investment for smallholders. But at the same time, it may induce more investment because of the demand for food. However, the immediate effect is negative due to the diversion of funds for consumption. Similarly, nonfarm income also increases investment variability because the farmers having additional income sources may stimulate more or less investment depending on their dependency on agriculture. Large families are constrained and face higher investment risk because of increased consumption expenditures, leading to a decline in investment (Joshi, 2004) and a rise in investment risk.

6. Conclusion and policy implications

Land ownership affects both on-farm investments and the ability to finance the investments. This paper investigated the determinants of on-farm investment and variability in on-farm investment in Indian agriculture using cross-sectional data from the Situational Assessment Survey of Agricultural Households (77th round, January 2019–December 2019) of NSSO. The empirical evidence from this study revealed that land ownership security positively and significantly affected on-farm investment among farmers in India. In addition, farm size, amount of credit, farmer’s age, access to irrigation, awareness of government programs, nonfarm income, rate of market orientation, and value of animal output significantly affected farmers’ on-farm investments. Indeed, the study found that land security and access to technical assistance, farmer education, assured irrigation, and market participation reduced on-farm investment risk (variability of on-farm investment). Perhaps lessons to policymakers from this study are to design policies that encourage farmers to borrow from formal sources and policies that increase technical assistance to farmers. To this end, banks and other formal credit providers could engage in educational campaigns and financial training for farmers. Institutional credit expansion likely will be needed to finance on-farm investment.

The main policy implication inferred is that the government should frame legal policies to strengthen the tenurial status of tenant farmers. A secure tenancy encouraging smallholders to increase farm investment and efficiency has already been evidenced in West Bengal through the policy of “Operation Barga” (Banerjee and Ghatak, 2004). So, in line with Operation Barga, there is a need for land tenancy reform in other regions of India to enhance agricultural productivity and production. Besides, based on the preceding analysis, we believe that concerted efforts are required to strengthen extension services and incentivize farmers to engage in training and awareness workshops regarding government agricultural policies. Furthermore, the resource-poor small and marginal farmers need to be provided with easy credit and brought within the network of crop insurance to protect them against the market and climatic risks. However, access to credit and crop insurance should be implemented appropriately, emphasizing the ease of getting compensation without any official hitch. On the other hand, the study revealed that household size negatively impacted on-farm investment in India.
The main limitation of the present study is that it has examined the issue of land tenure security and on-farm investment at the cross-sectional and aggregate level using pan-India data. However, a sub-national level analysis using the state or district level data would reveal the actual picture at the micro and meso levels. Investment risk may also be captured if we have time series data of many years. Technological progress, market conditions and various government policies also influence investment variability. We could not include all such variables due to a lack of suitable and relevant data. Future studies may try to fill those gaps.

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  Supervision: A.K., P.G.

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References


Appendix

Table A1. OLS results of determinants of land ownership status (1st stage)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>Std. Error</th>
<th>t-Statistic</th>
</tr>
</thead>
<tbody>
<tr>
<td>NCPH</td>
<td>−0.222</td>
<td>0.026</td>
<td>−8.618</td>
</tr>
<tr>
<td>Farm size</td>
<td>0.063</td>
<td>0.043</td>
<td>1.465</td>
</tr>
<tr>
<td>Credit accessed</td>
<td>0.233</td>
<td>0.026</td>
<td>8.948</td>
</tr>
<tr>
<td>Farmer’s age</td>
<td>0.007</td>
<td>0.010</td>
<td>0.701</td>
</tr>
<tr>
<td>Farmer’s education</td>
<td>0.035</td>
<td>0.004</td>
<td>8.911</td>
</tr>
<tr>
<td>Irrigation dummy</td>
<td>0.030</td>
<td>0.314</td>
<td>0.096</td>
</tr>
<tr>
<td>Technical assistance</td>
<td>1.017</td>
<td>0.245</td>
<td>4.177</td>
</tr>
<tr>
<td>Awareness about MSP</td>
<td>1.148</td>
<td>0.064</td>
<td>18.005</td>
</tr>
<tr>
<td>Nonfarm income</td>
<td>0.051</td>
<td>0.016</td>
<td>3.105</td>
</tr>
<tr>
<td>Household size</td>
<td>−0.145</td>
<td>0.051</td>
<td>−2.861</td>
</tr>
<tr>
<td>Market orientation</td>
<td>0.052</td>
<td>0.079</td>
<td>0.663</td>
</tr>
<tr>
<td>Crop insurance</td>
<td>0.168</td>
<td>0.098</td>
<td>1.708</td>
</tr>
<tr>
<td>Animal output</td>
<td>0.043</td>
<td>0.019</td>
<td>2.308</td>
</tr>
<tr>
<td>Constant</td>
<td>−0.093</td>
<td>0.384</td>
<td>−0.242</td>
</tr>
<tr>
<td>Adj. R-squared</td>
<td>0.681</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adjusted R-squared</td>
<td>0.611</td>
<td></td>
<td></td>
</tr>
<tr>
<td>F-statistic</td>
<td>13554.080</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Prob(F-statistic)</td>
<td>0.000</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: ***P < 0.01, **P < 0.05, *P < 0.10, MSP is the minimum support price of crops in India.
Source: Authors’ estimation from NSSO-SASAH data, 2019.
### Table A2. Determinants of farm investment (Second-stage results)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>Std. Error</th>
<th>Prob.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Land tenure security</td>
<td>0.230***</td>
<td>0.003</td>
<td>0.000</td>
</tr>
<tr>
<td>Farm size</td>
<td>0.251***</td>
<td>0.021</td>
<td>0.000</td>
</tr>
<tr>
<td>Credit accessed</td>
<td>0.801***</td>
<td>0.029</td>
<td>0.000</td>
</tr>
<tr>
<td>Farmer’s age</td>
<td>0.011***</td>
<td>0.001</td>
<td>0.000</td>
</tr>
<tr>
<td>Farmer’s education</td>
<td>0.120***</td>
<td>0.008</td>
<td>0.000</td>
</tr>
<tr>
<td>Irrigation dummy</td>
<td>0.209***</td>
<td>0.016</td>
<td>0.000</td>
</tr>
<tr>
<td>Technical assistance</td>
<td>0.579***</td>
<td>0.048</td>
<td>0.000</td>
</tr>
<tr>
<td>Awareness about MSP</td>
<td>0.129***</td>
<td>0.033</td>
<td>0.000</td>
</tr>
<tr>
<td>Nonfarm income</td>
<td>0.208***</td>
<td>0.021</td>
<td>0.000</td>
</tr>
<tr>
<td>Household size</td>
<td>−0.022**</td>
<td>0.007</td>
<td>0.001</td>
</tr>
<tr>
<td>Market orientation</td>
<td>0.158</td>
<td>1.528</td>
<td>0.907</td>
</tr>
<tr>
<td>Crop insurance</td>
<td>−0.016</td>
<td>0.117</td>
<td>0.890</td>
</tr>
<tr>
<td>Animal output</td>
<td>0.269***</td>
<td>0.038</td>
<td>0.000</td>
</tr>
<tr>
<td>Residual</td>
<td>0.381</td>
<td>0.733</td>
<td>0.603</td>
</tr>
<tr>
<td>Constant</td>
<td>0.621</td>
<td>0.603</td>
<td>0.310</td>
</tr>
<tr>
<td>Adjusted R-squared</td>
<td>0.696</td>
<td>F-statistic</td>
<td>16.99</td>
</tr>
<tr>
<td>Sum squared residuals</td>
<td>21034</td>
<td>Prob(F-stat)</td>
<td>0.000</td>
</tr>
</tbody>
</table>

Notes: ***P < 0.01, **P < 0.05, *P < 0.10. The standard error is bootstrapped standard error, as White’s robust standard errors donot correct for the use of generated regressors on the mean investment equationin the final stage. Source: Authors’ estimation from NSSO-SASAH data, 2019–21.