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Impact of land loss on academic performance among rural adolescents in China: based on cognition-investment-performance framework

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Introduction: As an exclusive group resulting from land requisition and demolition during the process of urbanization, the education of children belonging to landloss farming families has received worldwide attention. However, few studies have explored the mechanisms and effects of land loss on adolescents' academic performance.

Method: Using 5,133 family samples in 2014 and 3,810 family samples in 2018 from China Family Panel Studies (CFPS), this study employed the PSM-DID and KHB models to explore the impact of land loss on the academic performance of rural adolescents.

Results: The findings indicate that compared to non-land-loss families, the academic performances of adolescents in the land-loss families were lower. Additionally, land-loss families have a lower awareness of educational value. This verifies the logical mechanism of "land loss→ family education cognition→ family human capital investment→ adolescents' academic performance." Gender differences exist in the negative effects of land loss on adolescents' academic performance; land loss has a greater impact on boys.

Discussion: The government should strengthen the training system for land-loss farmers, improve the social security system and state of female-led families, and focus on boys affected by land loss.

KEYWORDS

land loss, academic performance, teenagers, gender difference, rural revitalization

1 Introduction

Adolescence is the most important phase in human capital formation (Heckman and Kautz, 2014), and the human capital of rural adolescents is pivotal in the sustainable development of the national economy and society. The academic performance of adolescents helps them realize their own human capital accumulation and evade the intergenerational transmission of poverty (Bai et al., 2019; Nong et al., 2022; Victora et al., 2022). However, the academic performance of urban and rural adolescents is still not at par (Wang et al., 2021; Liu and Helwig, 2022). According to the China Family Panel Studies (CFPS) 2020 data, only 20.46% of educated adolescents in rural families in China had "excellent" scores in both Chinese and mathematics. Therefore, there is an increasing awareness of the importance and

urgency of improving the academic performance of rural adolescents to promote balanced educational development (Agger et al., 2018; Boeren, 2019). The Chinese government has issued a series of documents, including the "No. 1 Central Committee Document of 2021" and "Opinions of the State Council on the Implementation of the Rural Revitalization Strategy," aimed at achieving the equalization of rural and urban education by prioritizing its development and improving the quality of education and teaching. However, despite the efforts of governments at all levels to promote the rural education guarantee mechanism, they still fail to effectively achieve the comprehensive and high-quality development of rural education.

In China, investment in education generally includes investment in both public education and family education. Compared with public government educational investment, family private educational investment has a greater influence on the resources and growth environment that enable children's development and adolescents' academic performance (Ghanney, 2018; Kim et al., 2020; Fretwell, 2021). Previous studies mainly focused on the relationship between parental involvement and children's academic performance, particularly discussing the influence of parents' educational expectations (Cross et al., 2019; Pinquart and Ebeling, 2020), personal characteristics (such as education and income levels) (Assari and Caldwell, 2019; Poon, 2020), personal status (such as social status and economic status) (Duan et al., 2018; Tan et al., 2020), and specific behaviors (such as parenting style and parent-child interaction) (Talin et al., 2021; Toor, 2021) on children's academic performance. These studies are useful for understanding the relationship between parental involvement and children's academic performance. However, research on the structural factors, such as parental background, influencing the mechanism of educational inequality is scant. With the comprehensive deepening of China's urban-rural integration process, a significant amount of rural collective land has been converted to urban construction land, resulting in a rapidly growing number of land-loss farmers as an exclusive group (Xie, 2019; Wang et al., 2020; Bao et al., 2021). Consequently, in terms of the right and quality of learning, the academic performance of Chinese land-loss farmers' adolescent children is adversely affected, along with the process of achieving balanced educational development.

Among the studies on the impact of land loss on the academic performance of adolescents, some studies have explored the inequality of educational opportunities between land-loss farmers and urban families based on time (Tang and Li, 2021). Additionally, some scholars have used social conflict patterns to analyze the impact of land acquisition on children's education (Le and Nguyen, 2020). However, this literature is based on the direct impact of land loss on children's education, while neglecting the invisible link between land loss and children's academic performance. Therefore, to fill this gap, the current study investigates the influencing variables of adolescents' academic performance based on the logical framework of "land loss→ family education cognition→ family human capital investment→ adolescents' academic performance of rural adolescents.

As a major life event, land loss has a significant effect on farmers' life choices. It leads to a loss of permanent income from land for

rural families (Li et al., 2018; Coulibaly and Li, 2020; Tuan, 2021), making it rational for them to seek non-farm work in cities. However, this often leads to frequent unemployment and unstable employment (Liu, 2020; Kang and Li, 2022), increasing the mental pressure on parents. This can result in them prioritizing their own livelihood decisions over those of their adolescent children's in terms of quality of learning motivation and learning rights, owing to their social status and emotional changes (Xu, 2020; Palit, 2022), and this pressure forces land-lost farmers to reduce expenditures on their adolescents' education. In addition, the mental stress caused by the loss of land may cause land-lost farmers to adopt inappropriate parenting methods (McLeod and Shanahan, 1993) and the intergenerational transfer of family culture can result in internalization of adolescents' learning attitude and academic expectations, thus affecting family education cognition and affecting their children's educational views (Yang, 2021; Furukawa Marques and Lagier, 2022). Land loss may affect adolescents' academic performance through parenting methods (education cognition) and human capital investment of the families.

In summary, this study focuses on identifying the impact of land loss on the academic performance of rural adolescents and analyzes the heterogeneous effects of differences in the gender of decision makers and the gender of adolescents. The innovative aspect of this study is as follows: 1) It tests the influence of the mechanism of land loss on rural adolescents' academic performance based on the adolescent stage of human capital investment, according to the logic of "land loss \rightarrow family education cognition \rightarrow family human capital investment → adolescents' academic performance; " 2) To improve the robustness of the research results, this study improves the covariates by setting the family characteristic variable, the household economic variable, and the community variables, while progressively enriching the dependent variable in terms of educational cognition, family human capital investment, and academic performance; 3) It thoroughly investigates the impact of land loss on the academic performance of adolescents based on the gender of decision makers and adolescents' gender differences and analyzes the reality of "gender equality" in the education of adolescents in rural China. The research content and implications of this article are also applicable to developing countries with similar national conditions and resource endowments as China, intending to serve as a lesson to these countries, particularly for issues related to land-loss.

The remainder of this paper is organized as follows: Section 2 introduces the theoretical framework, whereas Section 3 introduces the data sources, variable selection, and model construction. Section 4 presents the empirical results. Section 5 provides an in-depth discussion of these results. The conclusions and related policy recommendations are presented in Section 6.

2 Theoretical framework and hypotheses

The role of parents in family education is crucial. Parents can shape children's values and behaviors through parent-child relationships and family interactions, ultimately impacting their learning achievement through educational expectations (Veas et al., 2019; Pinquart and Ebeling, 2020). However, urbanization

has disrupted the lifestyle of land-lost farmers, forcing them to expend most of their energy and time on finding and adapting to new work, which can have negative impacts on the mental state of family members (Ding et al., 2020; Wu and Wang, 2021; Han et al., 2022). As a result, land-lost farmers equate children's education with formal schooling, and thus, neither devote time nor energy to it. Moreover, the education level of land-lost farmers is generally low, and they often lack the necessary skills to engage in non-agricultural industries, which limits their job opportunities during urban expansion (Liu et al., 2018; Memon et al., 2019). Their education level and employment environment limitations lead to the choice of livelihood strategies and their educational concepts and behaviors are limited by the cultural concept of "content with the *status quo*." This leads to a misestimation of the urban-rural integration process of education and neglecting the adolescents' education.

In terms of research methods, these studies have mostly focused on qualitative descriptions of cases in specific regions and lack empirical analyses based on nationally representative sample data in China. Although some studies have used Tobit, Tobit-IV, and DID models to explore the effects of court environment on adolescents' academic performance, they have been unable to well attenuate the effects of unobservable intergroup differences on assessment results. In terms of research content, while these studies have demonstrated an implicit link between landlessness and adolescents' human capital, they have not thoroughly examined the impact of landlessness on academic performance indicators that directly reflect individual human capital status and require a careful analysis of the causal mechanisms involved. Land acquisition can be considered the loss of the most important capital for livelihood (Xu et al., 2019; Le and Nguyen, 2020). Limited by resource endowment, land-lost farmers struggle to secure suitable employment in cities, leading to changes in their family's main income patterns. This difficulty in compensating for the loss of direct income through capital or labor after the loss of land often results in a decline in family income (Belay and Mengistu, 2019; Kojin, 2020). Given that family income significantly affects education consumption expenditure (Jabar et al., 2021; Wei et al., 2021), land-lost farmers struggle to afford the cost of their children's education when family economic conditions are severely constrained. As a result, children from many land-lost families fail to enter high-quality schools, negatively impacting their academic performance due to lack of investment in their education, as compared to those from non-land-lost families. Therefore, we propose the following hypotheses:

H1: The academic performance of adolescents from land-loss families is worse, and earning females have a promoting effect on the academic performance of children.

H2: Compared with non-land-loss rural families, land-loss farmers have lower levels of family education cognition and lower human capital investment.

The emotional and mental impact of land loss on farmers (Duncan et al., 2019; Chen, 2020) hinders their ability to guide their children, resulting in neglect of their children's studies and negatively impacting their academic performance. According to the family division of labor theory, housework and childcare responsibilities are exclusive to women (Koster et al., 2022).

Among land-lost farmers, female employment can provide additional economic support to the family and alleviate the economic pressure caused by land loss. Moreover, compared to fathers, mothers are more likely to allocate family resources to children's education and related expenditures (Yunxia and Xinrong, 2020; Wang and Cheng, 2021; Koster et al., 2022), which can improve their adolescents' academic performance. Additionally, boys are more competitive and sensitive to their environment than girls (Bully et al., 2019; Shi et al., 2021). In a land-lost environment, boys are more affected by the nonlearning-oriented surroundings and their performance is lower compared to girls. Therefore, we propose the following hypothesis:

H3: Among the children, the academic performance of boys is affected more by land loss than that of girls.

Figure 1 illustrates the pathways by which land loss affects adolescent academic performance.

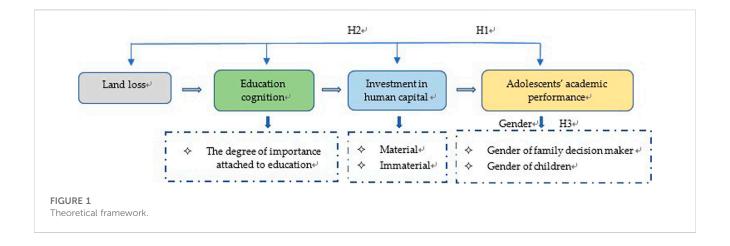
3 Data, variables, and methods

3.1 Data

The data used in this study were obtained from the 2014 and 2018 CFPS databases established by the China Center for Social Sciences Investigation (ISSS) of Peking University. The sample covered 25 provinces, municipalities, or autonomous regions. All family members identified in the 2010 baseline survey and their future biological or adopted children were permanently tracked as genetic members of the CFPS. Four types of questionnaires were used: community, family, adult, and children, and the survey objects included all family members in the sample households. The land situation and human capital investment status of the interviewed households was investigated in detail. The study focused on rural families with adolescent children aged 10-15 and excluded families with incomplete land acquisition information or without schoolaged children. After data cleaning, there were 5,133 household samples in 2014, with 312 belonging to the treatment group (land was expropriated before 2018) and the remaining 4,821 in the control group. In 2018, the sample consisted of 3,810 households, with 295 belonging to the treatment group (land acquired during the current period), and the remaining 3,515 in the control group.

3.2 Variables

The aim of this study was to investigate the impact and mechanism of land loss on the academic performance of rural adolescents. The core treatment variable was *treated*, which denoted whether land was acquired. Before applying the PSM-DID model, the treatment variable was constructed based on whether the rural family experienced land acquisition in 2018 and not 2014. If so, it was included in the treatment group with a value of 1. If a rural household did not experience land expropriation in both 2014 and 2018, it was set in the control group with a value of 0. According to Table 1,



6.78% of the rural household land in the sample was expropriated.

Following the analytical logic of "land loss→ family education cognition→ family human capital investment→ adolescents' academic performance", the dependent variables in this study were divided into three levels: 1) *childsave* and *saving*, which represented the family's emphasis on education, for analyzing the impact of land loss on family education cognition; 2) *spend* and *talk* for measuring human capital investment in education expenditure and non-material human capital investment in children's learning, respectively; 3) *class rank*, *grade rank*, *Chinese*, and *math*, which represented the adolescents' academic performance, for analyzing the effects of land loss on the accumulation of adolescent human capital, and on the relative and absolute evaluation of academic performance.

Multiple factors can affect adolescents' academic performance. In addition to the key variable of land acquisition, this study selected covariates based on the literature (Zhihua et al., 2015; Junlong, 2017; Haochen, 2019). We included 13 indicators, including family characteristics, family economy, and community variables. Table 1 presents the variables and descriptive statistics used in this study.

3.3 Methods

Y was set as the outcome variable (academic performance of adolescents) influenced by the loss of land. Intuitively, the effect of land loss on the academic performance of adolescents from rural families is as follows:

$$ATT = Y_{t1} - Y_{t0}, (1)$$

where ATT represents the effect of land loss on rural families (treatment effect), Y_{t0} represents the outcome variable values of families in the treatment group in 2014, and Y_{t1} represents the 2018 outcome variable values for households in the treatment group. However, Eq. 1 may be biased because even if the land is not expropriated, the value of Y may change with time; that is, Eq. 1 ignores the change of time trend. Thus, Eq. 2 is used to correct this bias:

$$ATT = (Y_{t1} - Y_{t0}) - (Y_{t1f} - Y_{t0}), \tag{2}$$

where Y_{t1f} represents the outcome variable in 2018 that could not be observed if the treatment group had not undergone land acquisition, and $(Y_{t1f} - Y_{t0})$ represents the change in the academic performance of adolescents in the treatment group without land acquisition, which is not observable.

Assuming that the treatment and control groups have the same time trend, Eq. 3 is used for DID estimation:

$$ATT = (Y_{t1f} - Y_{t0}) - (Y_{c1} - Y_{c0}), \tag{3}$$

where Y_{c1} represents the outcome variable in the control group in 2018, and Y_{c0} represents the outcome variable of the control group in 2014. Eq. 3 is a common DID estimation method that must satisfy the common trend assumption. The validity of the estimation depends on whether $(Y_{t1f} - Y_{t0})$ and $(Y_{c1} - Y_{c0})$ are equal.

PSM is a commonly used method to study land issues, but PSM-DID has the advantage of further attenuating the effect of unobservable intergroup differences on assessment results compared to PSM alone. If the assumption of a common trend is not satisfied, "selection bias" of the samples will occur, and the reliability of the model results will be compromised. PSM constructs counterfactual events by finding samples from the control group that have similar characteristics to those in the treatment group. It thereby overcomes "selection bias," and by combining PSM with the DID method, we can control for non-observable, time-dependent between-group differences, and address endogeneity. Overall, PSM-DID can reduce or eliminate the difference in time trends between the two types of households by estimating propensity scores for dislocated and non-dislocated households through observable covariates and matching each dislocated household with its closest non-dislocated household, solving the endogeneity problem caused by sample selection. In this study, we adopted the PSM-DID method to effectively identify and evaluate the "treatment effect" using STATA 17.0 software (Fan and Zhang, 2021; Zhang et al., 2022).

This study implemented the differential PSM model proposed by Heckman et al. (1998) to investigate the impact of land loss on rural families' human capital investment and adolescents' academic performance. Prior to estimation, the two types of families were matched based on propensity scores. Specifically, propensity scores

TABLE 1 Variable selection and descriptive statistical analysis.

Variables		Name	Description	Mean	Std. dev	Min	Max
Trea	Treatment variable		Did you go through eminent domain: yes = 1, no = 0	0.068	0.251	0	
Covariates	Family characteristic variables	agep	Age of the household head	35.396	8.477	20	84
		agep2	Square of the age of the head of the household	1,324.728	679.560	400	7,056
		gender	Gender of the household head: female = 0 , male = 1	0.496	0.500	0	1
		edumax	Highest level of education in the family: illiteracy/semi-illiteracy = 1, primary = 2, junior high = 3, senior high = 4, junior college = 5, bachelor's degree = 6, master's degree = 7, doctorate = 8 ± 7	3.124	1.556	1	8
		fs	Number of books in the home: none = 0, 1-10 books = 1, 11-20 books = 2, 21-50 = 3, 51-100 = 4, 101-500 = 5, 501-1,000 = 6, more than 1,000 = 7	1.715	1.761	0	7
		young	Number of adolescents aged 10-15 years	1.524	0.696	1	6
		old	Number of family members aged 60 years and above	0.846	0.901	0	2
	Household economic variables	fincome1_per	Per capita net household income (yuan)	10,990.170	11,040.840	0	168,625
		total_asset	Household net worth (yuan)	307,931.200	916,488.600	-704425	50,000,000
	$\label{eq:community} \begin{tabular}{lll} fm & Whether anyone is self-employed: yes = 1, \\ no = 0 & \\ \begin{tabular}{lll} Community variables & cg & Time to provincial capital (hours), the most common mode of transportation used by people from the villages to go to the provincial capital of the province, such as by car, train, plane, etc. \\ \begin{tabular}{lll} Plane & Plane &$			0.094	0.292	0	1
			5.408	5.857	0	60	
		ch	Proportion of migrant workers (%)	37.852	22.209	0	90
		provcd	eastern region = 1, central region = 2, western region = 3	2.073	0.837	1	3
Dependent variables	Cognition of family education	childsave	Whether you save money for children's education: no = 0, yes = 1	0.540	0.500	0	1
		saving	Money saved for children's education in the past 12 months (yuan)	8,882.285	13,841.060	0	200,000
	Household investment in human capital	talk	How often you talk to your child about school: never = 1, rarely (once a month) = 2, occasionally (once a week) = 3, often (2–4 times a week) = 4, often (5–7 times a week) = 5	2.985	1.207	1	5
		spend	Total Expenditure on education in the past 12 months (yuan)	2,925.297	3,622.548	0	36,300
	Academic performance of adolescents	class rank	Last midterm and final exam class ranking (%): top 10% = 1, 11%-25% = 2, 26%-50% = 3, 51%-75% = 4, bottom 24% = 5	2.804	1.696	1	5
		graderank	Last midterm and final exam grade ranking (%): top 10% = 1, 11%-25% = 2, 26%-50% = 3, 51%-75% = 4, bottom 24% = 5	3.216	1.850	1	5
		math	Math score: poor = 1, medium = 2, good = 3, excellent = 4	2.293	1.048	1	4
		chinese	Chinese score: poor = 1, medium = 2, good = 3, excellent = 4	2.266	0.992	1	4

were calculated for land-loss and non-land-loss households based on observable household characteristics (covariates). Then, each land-loss household was matched with the closest non-land-loss household, thereby eliminating the time trend of two types of families. Subsequently, the treatment effect was calculated using the DID method as per Eq. 3. The specific steps taken are as follows:

First, PSM processing was performed. A logit model was constructed for regression with *treated* as the explained variable, and a group of new observed samples were obtained by "1:1 nearest neighbor" matching method.

$$logit(treated = 1) = \alpha_0 + X_{it}\beta + \varepsilon_{it}, \tag{4}$$

where Treated indicated whether rural households were treated by land acquisition, if so treated = 1, else treated = 0. X_{it} represented the covariate, including the family characteristic variable, family economic variable, and community variable. ε_{it} was the residual.

Second, the difference model was constructed as:

$$Y_{it} = \alpha + \gamma treated + X_{it}\delta + \mu_i + \vartheta_t + \varepsilon_{it}, \qquad (5)$$

where γ represented the treatment effect of the policy, X_{it} was the covariate, μ_i was an individual fixed effect, ϑ_t was a time fixed effect, and ε_{it} was the residual term. The dependent variables were divided into three levels: family education cognition, human capital investment and adolescents' academic performance. The hypotheses will be tested sequentially in the empirical analysis.

To identify the mechanism of land loss on the academic performance, this paper draws on Breen et al. (2013) and uses the KHB mediation effects test to test whether education cognition and family human capital investments affect adolescents' academic performance. For the test of mediating effects, the traditional test of mediating effects is only applicable to linear models and cannot test nonlinear models. The KHB mediation effect test can effectively decompose linear and nonlinear regression models of the mediating effects.

4 Results

4.1 The influence of land loss on the academic performance of adolescents

4.1.1 Test of balance

An important prerequisite for the application of the PSM-DID method is to balance the control variables. The balancing hypothesis requires that the bias between the matched treatment and control groups is less than 5%, or that the *t*-test results show no significant difference between the matched treatment and control groups. Table 2 presents the balance test results for the control variables before and after matching. The mean standard error of the control variables decreased from 16.1% to 3.8% after matching. The Pseudo R2 value, which measures the goodness of fit of propensity score regression, was low after matching. The *p*-value of the joint significance test of the coefficients of the control variables was 0.994, indicating that the coefficients of the control variables were jointly significant before matching, and the

null hypothesis that the coefficients of the control variables were jointly 0 could not be rejected after matching. To ensure the quality of matching between samples, the kernel density plots were further plotted after obtaining the propensity scores to examine the common support domain after sample matching (see Figure 2). It can be seen that there is a large range of overlap between the propensity scores of the experimental and control groups after matching, and most of the observations are in the common range of values. The above test results demonstrate that this study has well-matched land-lost families and non-land-lost families, the common support assumption is satisfied, so the PSM-DID model is applicable to this study.

4.1.2 Benchmark regression results

Table 3 presents the results of land loss on the academic performance of rural adolescents. Models 1 to 4 use class rank, grade rank, Chinese score, and math score, respectively, as the dependent variables, and include land-loss variables and covariates. The first two columns analyze the academic performance differences between land-loss families and nonland-loss families, while the last two columns show the differences in single subject scores of adolescents. According to Models 1 and 2, after controlling for other variables, the landloss variables are all significantly positive at the 5% significance level, suggesting that the achievement ranking of adolescents from land-loss families is lower than that of non-land-loss families. Model 3 shows that the coefficient of the land-loss variable is -0.7352, which is significantly negative, indicating that compared to the non-land-loss families, the Chinese scores of the adolescents from land-loss families are lower. Model 4 shows similar results, with the math scores of the land-loss families being lower than those of the non-land-loss families. Furthermore, combining Models 3 and 4, it can be concluded that loss of land has a greater impact on the performance in Chinese of adolescents in rural families. Therefore, Hypothesis 1 is confirmed, suggesting that when rural families face the high external risk of land expropriation, the physical health and mental states of parents are affected, and this negative effect is transmitted across generations, leading to a reduction in academic achievements and performance of adolescent girls.

4.2 Mechanism analysis

4.2.1 The influence of land loss on the family value cognition

According to the analytical logic of "land loss→ family education cognition→ family human capital investment→ adolescents' academic performance," Table 4 presents the results of the PSM-DID model to assess the impact of land loss on rural family education cognition (the degree of emphasis on adolescents' education). The coefficients of land-loss variables in Models 5 and 6 are −0.3516 and −0.2118, respectively. This indicates that land-loss families have a lower awareness of saving for their children, and the amount of money saved is significantly lower than that of non-land-loss families, which implies that the loss of land significantly reduces the cognition of the educational value for rural families.

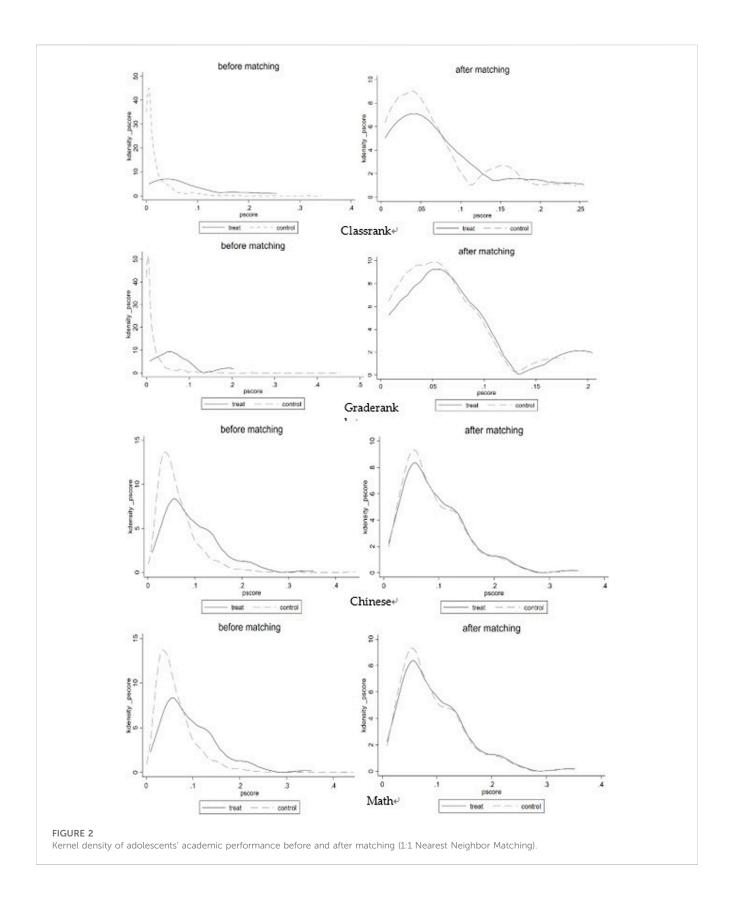
TABLE 2 Balance test results of control variables before and after matching.

gender U 0.496 0.495 0.200 0.030 0.974 agep U 36.130 35.414 8.200 1.340 0.181 agep2 U 1.36.130 35.414 8.200 1.340 0.181 agep2 U 1.383.900 1.326.000 8.300 1.340 0.774 chumax U 2.943 3.108 -11.100 -0.030 0.974 chumax U 2.943 3.108 -11.1300 -0.030 0.974 fs U 2.7385 13.259 2.6700 4.620 0.000 fs U 2.7385 13.259 2.6700 4.620 0.000 g U 1.584 1.531 6.700 1.210 0.228 poung U 1.584 1.525 7.400 0.630 0.409 old U 0.859 0.861 -0.200 0.874 lmincomet_per U 9.457 9.458	Variable	Unmatched matched	Mean treated	Mean control	Bias%	T-value	<i>p</i> -value
agep U 36,130 35,414 8,200 1,349 0,181 M 36,130 36,342 -2,400 -0,270 0,789 agep2 U 1,383,900 1,326,400 8,300 1,340 0,180 edumax U 2,943 3,108 -11,300 -0,340 0,734 fb U 2,943 3,036 -6,400 -0,360 0,723 fb U 27,385 13,229 26,700 46,20 0,000 young U 1,584 1,531 6,700 1,210 0,228 young U 1,584 1,531 6,700 1,210 0,228 M 1,584 1,531 6,700 1,210 0,228 M 0,859 0,887 -3,100 -0,360 0,723 Infincomet_per U 9,457 8,867 61,700 8,880 0,000 Infincomet_per U 1,2431 12,691 32,100 51,29	gender	U	0.496	0.495	0.200	0.030	0.972
M		M	0.496	0.508	-2.300	-0.260	0.794
agep2 U 1,383,900 1,326,400 8.300 1,340 0,734 edumax U 2,943 3,108 -1,100 -0,300 0,734 fs U 2,943 3,108 -1,130 -0,030 0,723 fs U 2,7385 13,259 26,700 46,20 0,000 young U 1,584 1,531 6,700 12,10 0,228 dd U 1,584 1,531 6,700 12,10 0,228 dd U 0,859 0,861 -0,200 -0,30 0,723 lnfincomel_per U 9,457 8,867 61,700 8,880 0,000 lntotal_asset U 12,431 12,941 -1,100 -0,130 0,984 fm U 1,2431 12,941 -1,100 -0,130 0,895 fm U 1,12,31 12,941 -1,100 -0,130 0,895 fm U 1,2431	agep	U	36.130	35.414	8.200	1.340	0.181
edumax M 1,383,900 1,405,600 -3,100 -0,340 0,744 edumax U 2,943 3,108 -11,300 -0,030 0,774 M 2,943 3,036 -6,400 -0,360 0,723 fs U 27,385 13,259 26,700 4,620 0,000 young U 1,584 1,531 6,700 1,210 0,228 dd U 0,839 0,861 -0,200 -0,300 0,723 lnfincomel_per U 9,457 8,867 61,700 8,880 0,000 lntotal_asset U 12,431 12,941 -1,100 -0,300 0,734 fm U 1,2431 12,941 -1,100 -0,300 0,784 fm U 1,12431 12,941 -1,100 -0,300 0,885 fm U 1,12431 12,444 -1,100 -0,300 0,885 fg U 2,130 0,895		M	36.130	36.342	-2.400	-0.270	0.789
edumax U 2.943 3.108 -11.300 -0.030 0.974 M 2.943 3.036 -6.400 -0.360 0.723 fs U 27.385 13.259 26.700 4.620 0.000 young U 1.584 1.531 6.700 1.210 0.228 M 1.584 1.531 6.700 1.210 0.228 M 1.584 1.525 7.400 0.830 0.409 M 0.889 0.861 -0.200 -0.30 0.974 M 0.889 0.861 -0.200 -0.30 0.723 Infincomet_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 -0.100 -0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 fm U 0.130 0.869 13.800 2.410 0.016 eg	agep2	U	1,383.900	1,326.400	8.300	1.340	0.180
In M 2.943 3.036 −6.400 −0.360 0.723 Is U 27.385 13.259 26.700 4.620 0.000 M 27.385 34.261 −13.000 −0.850 0.397 young U 1.584 1.531 6.700 1.210 0.228 M 1.584 1.525 7.400 0.830 0.409 old U 0.859 0.861 −0.200 −0.030 0.974 M 0.859 0.887 −3.100 −0.360 0.723 Infincome1_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 −0.100 −0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 fm U 0.130 0.869 13.800 2.410 0.016 cg U 25.107 23.957 13.200 2.070 0.38		M	1,383.900	1,405.800	-3.100	-0.340	0.734
fs U 27.385 13.259 26.700 4.620 0.000 M 27.385 34.261 −13.000 −0.859 0.397 young U 1.584 1.531 6.700 1.210 0.228 M 1.584 1.525 7.400 0.830 0.409 old U 0.859 0.861 −0.200 −0.030 0.974 M 0.859 0.887 −3.100 −0.360 0.723 Infincome1_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 −0.100 −0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.444 −1.100 −0.130 0.885 fm U 0.130 0.869 13.800 2.410 0.016 cg U 25.107 23.957 13.200 2.070 0.038	edumax	U	2.943	3.108	-11.300	-0.030	0.974
young U 1.584 1.531 6.700 1.210 0.228 young U 1.584 1.531 6.700 1.210 0.228 M 1.584 1.525 7.400 0.830 0.409 old U 0.859 0.861 -0.200 -0.030 0.723 Infincemel_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 -0.100 -0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.244 -1.100 -0.130 0.895 fm U 0.130 0.869 13.800 2.410 0.016 cg U 25.107 23.957 13.200 2.070 0.388 ch U 17.500 19.722 -22.700 -3.460 0.001 m 17.500 17.582 -0.800 -0.100 0.923		M	2.943	3.036	-6.400	-0.360	0.723
young U 1.584 1.531 6.700 1.210 0.228 old U 0.859 0.861 -0.200 -0.030 0.974 Infincome1_per U 0.859 0.887 -3.100 -0.360 0.723 Infincome1_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 -0.100 -0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 fm U 0.130 0.869 13.800 2.410 0.016 Gg U 0.130 0.869 13.800 2.410 0.016 Gg U 25.107 23.957 13.200 2.070 0.038 Gg U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 Proved U 2.130 2.193 -7.400 </td <td>fs</td> <td>U</td> <td>27.385</td> <td>13.259</td> <td>26.700</td> <td>4.620</td> <td>0.000</td>	fs	U	27.385	13.259	26.700	4.620	0.000
old M 1.584 1.525 7.400 0.830 0.409 old U 0.859 0.861 −0.200 −0.300 0.974 M 0.859 0.887 −3.100 −0.360 0.723 Infincome1_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 −0.100 −0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.444 −1.100 −0.130 0.895 fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 eg U 25.107 23.957 13.200 2.070 0.038 ch U 17.500 19.722 −22.700 −3.460 0.001 proved U 2.130 2.091 4.500 0.740 0.461		M	27.385	34.261	-13.000	-0.850	0.397
old U 0.859 0.861 -0.200 -0.030 0.974 M 0.859 0.887 -3.100 -0.360 0.723 Infincmet_per U 9.457 8.867 61.700 8.880 0.000 M 9.457 9.458 -0.100 -0.020 0.984 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.444 -1.100 -0.130 0.895 fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 eg U 25.107 23.957 13.200 2.070 0.038 ch U 17.500 19.722 -22.700 -3.460 0.001 proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387	young	U	1.584	1.531	6.700	1.210	0.228
Infincome1_per U 9.457 8.867 61.700 8.880 0.000 Intotal_asset U 9.457 8.867 61.700 8.880 0.000 Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.444 −1.100 −0.130 0.895 fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 cg U 25.107 23.957 13.200 2.070 0.038 ch U 17.500 19.722 −22.700 −3.460 0.001 proved U 2.130 2.091 4.500 0.740 0.461 proved U 0.068 — — — — Pseudo R2 U 0.068 — — — — M 0.005 — — — — — <td></td> <td>M</td> <td>1.584</td> <td>1.525</td> <td>7.400</td> <td>0.830</td> <td>0.409</td>		M	1.584	1.525	7.400	0.830	0.409
Infincomel_per	old	U	0.859	0.861	-0.200	-0.030	0.974
M		M	0.859	0.887	-3.100	-0.360	0.723
Intotal_asset U 12.431 12.091 32.100 5.120 0.000 M 12.431 12.444 −1.100 −0.130 0.895 fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 cg U 25.107 23.957 13.200 2.070 0.038 M 25.107 25.147 −0.500 −0.050 0.958 ch U 17.500 19.722 −22.700 −3.460 0.001 M 17.500 17.582 −0.800 −0.100 0.923 proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 −7.400 −0.870 0.387 Pseudo R2 U 0.068 − − − − M 0.005 − − − − M 3.640 −	lnfincome1_per	U	9.457	8.867	61.700	8.880	0.000
fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 cg U 25.107 23.957 13.200 2.070 0.038 Ch U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 Proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - M 0.005 - - - - - - M 3.640 - - - - - M 3.640 - - - - - M 3.640 - - - - - M 3.640 - </td <td></td> <td>M</td> <td>9.457</td> <td>9.458</td> <td>-0.100</td> <td>-0.020</td> <td>0.984</td>		M	9.457	9.458	-0.100	-0.020	0.984
fm U 0.130 0.869 13.800 2.410 0.016 M 0.130 0.1260 1.200 0.130 0.896 cg U 25.107 23.957 13.200 2.070 0.038 M 25.107 25.147 -0.500 -0.050 0.958 ch U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - M 0.005 - - - - - M 3.640 - - - - - M 3.640 - - - - - M 3.640 - - - - - M 0.000 - -	Intotal_asset	U	12.431	12.091	32.100	5.120	0.000
cg		M	12.431	12.444	-1.100	-0.130	0.895
cg U 25.107 23.957 13.200 2.070 0.038 M 25.107 25.147 -0.500 -0.050 0.958 ch U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - M 0.005 - - - - - LR chi² U 156.130 - - - - - M 3.640 - - - - - P > chi2 U 0.000 - - - - -	fm	U	0.130	0.869	13.800	2.410	0.016
M 25.107 25.147 -0.500 -0.050 0.958 ch U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 Proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - M 0.005 - - - - - LR chi² U 156.130 - - - - - M 3.640 - - - - - P > chi2 U 0.000 - - - - -		M	0.130	0.1260	1.200	0.130	0.896
ch U 17.500 19.722 -22.700 -3.460 0.001 M 17.500 17.582 -0.800 -0.100 0.923 proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - M 0.005 - - - - - LR chi² U 156.130 - - - - - M 3.640 - - - - - P > chi2 U 0.000 - - - - -	cg	U	25.107	23.957	13.200	2.070	0.038
M 17.500 17.582 -0.800 -0.100 0.923 provcd U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - M 0.005 - - - - LR chi² U 156.130 - - - - M 3.640 - - - - - P > chi2 U 0.000 - - - - -		M	25.107	25.147	-0.500	-0.050	0.958
proved U 2.130 2.091 4.500 0.740 0.461 M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 - - - - - - M 0.005 - - - - - - LR chi² U 156.130 - - - - - M 3.640 - - - - - - P > chi2 U 0.000 - - - - -	ch	U	17.500	19.722	-22.700	-3.460	0.001
M 2.130 2.193 -7.400 -0.870 0.387 Pseudo R2 U 0.068 M 0.005 LR chi² U 156.130 M 3.640 P > chi2 U 0.000		M	17.500	17.582	-0.800	-0.100	0.923
Pseudo R2 U 0.068 — — — M 0.005 — — — — LR chi² U 156.130 — — — — M 3.640 — — — — P > chi2 U 0.000 — — — —	provcd	U	2.130	2.091	4.500	0.740	0.461
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		M	2.130	2.193	-7.400	-0.870	0.387
LR chi ² U 156.130 — — — — — — — — — — — — — — — — — —	Pseudo R2	U	0.068	_	_	_	_
M 3.640 — — — — — — — P > chi2 U 0.000 — — — — — —		M	0.005	_	_	_	_
P > chi2 U 0.000 — — — —	LR chi ²	U	156.130	_	_	_	_
V 2004		M	3.640	_	_	_	_
M 0.994 — — — —	P > chi2	U	0.000	_	_	_	_
		M	0.994	_	_	_	_

Possibly, rural families are devastated after the loss of land, and their living standards significantly decrease. Rural families may prioritize their survival and may be unable to devote enough time and energy to children to alleviate material hardship, which has negative consequences for child development (Brooks-Gunn et al., 2013; Desmond and Kimbro, 2015).

4.2.2 The impact of land loss on the family human capital investment

Table 5 presents the PSM-DID estimation results of the impact of land loss on the human capital investment of rural families. In Model 7, the coefficient of land-loss variable is 0.668, which is significant at the 1% level, indicating that compared to families whose land is not expropriated, families who have lost



their land talk less frequently to their children about school and pay less attention to their children's school conditions. This is because rural families face external risks due to land expropriation and must choose a new livelihood, causing

parents to devote more time and energy to employment and relax their focus on their children's study requirements and discipline. In Model 8, the variable of land loss is significantly negative. After land loss, families' adolescent education

TABLE 3 Influence of land-loss on the academic performance of adolescents.

Classrank Chinese -0.735*** treated 2.610** 2.888** -0.675** (2.444)(2.359)(-2.390)(-2.611)gender -0.835*** -0.964*** -0.787*** -0.810*** (-3.975)(-4.205)(-7.323)(-7.521)1.336*** 1.349*** 1.047*** 1.054*** agep (16.193)(16.031) (27.375) (27.503)-0.013*** -0.013*** -0.010*** -0.010*** agep2 (-14.396)(-14.200)(-22.520)(-22.585)-0.010-0.004-0.010-0.004cg (-0.833)(-0.789)(-0.710)(-0.661)-0.015-0.019*0.002 0.001 ch (-1.359)(-1.683)(0.339)(0.188)lnfincome1_per 0.478*** 0.531*** 0.225*** 0.222*** (2.721)(2.955)(2.788)(2.743)Intotal asset 0.209* 0.128 -0.171*** -0.172*** (1.842)(0.973)(-3.025)(-3.049)-0.692** -0.729** fm -0.118-0.112(-2.082)(-0.595)(-2.051)(-0.627)fs 0.018*** 0.017*** 0.006*** 0.005*** (6.214) (6.834)(4.942)(4.538)0.221*** 0.079 0.042 0.186** young (0.465)(0.230)(2.360)(2.792)0.489*** 0.549*** 0.531*** 0.538*** old (3.707)(3.914)(7.281)(7.367)edumax -0.259*** -0.217** -0.119*** -0.113*** (-2.778)(-2.049)(-3.262)(-3.091)-0.350*** -0.388*** -0.157** provcd -0.136** (-2.591)(-2.694)(-2.046)(-2.354)0.231 0.246 0.269 0.271 Adj. R2

Note: t value in parentheses; ***p < 0.01 **p < 0.05 *p < 0.1.

expenditure decreases, which is 0.22 units lower than that of families without land loss. These results are consistent with the findings in the literature (Leventhal and Newman, 2010; Brooks-Gunn et al., 2013; Desmond and Kimbro, 2015). Therefore, Hypothesis 2 is thus confirmed.

4.2.3 Land loss, human capital investment, and adolescent academic performance

To further verify the existence of a behavior-shaping mechanism between land-loss status and adolescents' academic performance,

TABLE 4 Impact of land loss on the family education cognition.

Variables	(5)	(6)
	Childsave	Lnsaving
treated	-0.352***	-0.212**
	(-11.117)	(-2.396)
gender	-0.022*	0.063
	(-1.829)	(0.737)
agep	0.009**	-0.048*
	(2.024)	(-1.705)
agep2	-0.0001	0.0003
	(-1.552)	(1.052)
edumax	0.029***	0.027
	(7.052)	(0.814)
fs	0.002***	0.0007*
	(16.505)	(1.708)
young	0.045***	-0.057
	(5.078)	(-1.007)
old	-0.002	0.034
	(-0.246)	(0.583)
lnfincome1_per	0.118***	0.025
	(13.056)	(0.409)
lntotal_asset	0.038***	0.126***
	(6.095)	(2.758)
fm	-0.057***	0.262*
	(-2.693)	(1.911)
cg	-0.003***	-0.008
	(-4.669)	(-1.602)
ch	-0.003***	0.013***
	(-4.161)	(2.830)
provcd	0.030***	0.170***
	(4.014)	(3.211)
Adj. R ²	0.116	0.060

Note: t value in parentheses; *** p < 0.01 ** p < 0.05 * p < 0.1.

this section presents mechanism tests based on the KHB method (Breen et al., 2013) following the logic of "land loss→ family education cognition →family human capital investment→ adolescents' academic performance". Models 8 and 9 (as presented in Table 6) verify the relationship between family education cognition and family human capital investment. Model 8 was a base model that contained only the control variables. Model 9 added a variable to measure the value of family education. The results of Model 9 show that after adding the variable of family education cognition, the coefficient of the land-loss variable

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TABLE 5 Impact of land loss on the household human capital investment.

Variables	(7)	(8)
	Talk	Lnspend
treated	0.668***	-0.222*
	(0.207)	(-1.953)
gender	-0.176*	0.102
	(0.103)	(0.928)
agep	0.335***	-0.028
	(0.047)	(-0.706)
agep2	-0.004***	0.0003
	(0.0006)	(0.734)
edumax	-0.005	0.137***
	(0.038)	(2.628)
fs	0.003***	-0.001
	(0.0007)	(-1.390)
young	0.108	-0.164**
	(0.069)	(-2.514)
old	0.092	0.005
	(0.066)	(0.073)
lnfincome1_per	0.287***	0.068
	(0.067)	(0.900)
lntotal_asset	0.081	0.073
	(0.052)	(1.324)
fm	-0.077	0.009
	(0.183)	(0.057)
cg	-0.008	0.002
	(0.006)	(0.238)
ch	-0.013**	-0.004
	(0.005)	(-0.617)
provcd	-0.079	-0.238***
	(0.063)	(-3.515)
Adj. R²	0.058	0.052

Note: t value in parentheses; ***p < 0.01 **p < 0.05 *p < 0.1.

decreases compared to the base model and is significant at the 5% level. This indicates that the negative effect of land loss on adolescents' academic performance can be alleviated by improving families' perception of educational value. Moreover, Models 1 and 10 were used to verify the relationship between family human capital investment and adolescents' academic performance. Model 10 included education expenditure to measure family human capital investment in comparison to the basic Model 1 with just the control variables. The results of Model 10 show that the coefficient of the land-loss variable

TABLE 6 Results of KHB mediated effects test.

Variables	(8)	(9)	(1)	(10)
	Lnspend	Lnspend	Classrank	Classrank
treated	-0.222*	-0.193**	2.610**	2.405**
	(-1.953)	(-1.985)	(2.444)	(2.025)
gender	0.102	0.022	-0.835***	-0.423
	(0.928)	(0.446)	(-3.975)	(0.317)
agep	-0.028	0.016	1.336***	0.994***
	(-0.706)	(0.738)	(16.193)	(0.164)
agep2	0.0003	-0.0002	-0.013***	-0.010***
	(0.734)	(-0.804)	(-14.396)	(0.002)
edumax	0.137***	0.063**	-0.259***	0.055
	(2.628)	(3.400)	(-2.778)	(0.150)
fs	-0.001	-0.000	0.018***	0.005*
	(-1.390)	(-0.092)	(6.834)	(0.002)
young	-0.164**	-0.132***	0.079	0.302
	(-2.514)	(-4.066)	(0.465)	(0.221)
old	0.005	0.002	0.489***	0.243
	(0.073)	(0.049)	(3.707)	(0.193)
lnfincome1_per	0.068	0.213***	0.478***	0.121
	(0.900)	(6.582)	(2.721)	(0.255)
lntotal_asset	0.073	0.009	0.209*	0.063
	(1.324)	(0.361)	(1.842)	(0.173)
fm	0.009	-0.058	-0.692**	0.414
	(0.057)	(-0.671)	(-2.051)	(0.593)
cg	0.002	0.000	-0.010	-0.006
	(0.238)	(0.007)	(-0.833)	(0.017)
ch	-0.004	0.001	-0.015	-0.037**
	(-0.617)	(0.559)	(-1.359)	(0.017)
provcd	-0.238***	-0.347***	-0.350***	-0.412**
	(-3.515)	(-11.623)	(-2.591)	(0.202)
lnsaving	_	0.078***	_	_
	_	(3.446)	_	_
lnspend	_	_	_	4.581***
	_	_	_	(0.399)
Adj. R ²	0.052	0.093	0.231	0.393

Note: t values are in parentheses; ***p < 0.01 **p < 0.05 *p < 0.1.

significantly decreases compared to that in Model 1, and the coefficient of *lnspend* variable is significantly positive, indicating that the negative pass-through effect of land loss on adolescents' academic performance can also be effectively compensated by the

TABLE 7 Gender heterogeneity of household decision makers.

Variables	Class	srank	Grad	erank	Chi	nese	Math	
	Man	Woman	Man	Woman	Man	Woman	Man	Woman
Treated	2.971***	2.438***	2.986*	2.529***	-1.013**	-0.845*	-0.932**	-0.429
	(2.154)	(2.361)	(1.914)	(0.359)	(-2.535)	(-1.712)	(-2.331)	(-1.078)
Family characteristic variables	controlled							
Household economic variables	controlled							
Community variables	controlled							

Note: t value in parentheses; ***p < 0.01 **p < 0.05 *p < 0.1.

TABLE 8 Gender differences of adolescents.

Variables	Class	rank	Grade	erank	Chir	Chinese Math		
	Воу	Girl	Воу	Girl	Воу	Girl	Воу	Girl
Treated	4.129***	1.566	4.051**	1.483	-1.114**	-0.797**	-0.695*	-0.670*
	(2.339)	(1.124)	(2.229)	(0.877)	(-2.251)	(-2.029)	(-1.715)	(-1.692)
Family characteristic variables	controlled	controlled	controlled	controlled	controlled	controlled	controlled	controlled
Household economic variables	controlled	controlled	controlled	controlled	controlled	controlled	controlled	controlled
Community variables	controlled	controlled	controlled	controlled	controlled	controlled	controlled	controlled

Note: t value in parentheses; *** p < 0.01 ** p < 0.05 * p < 0.1.

increase in human capital investment. It can be preliminarily concluded that the academic achievement difference of adolescents from land-loss families is affected by their cognition of educational value, the scale of educational resources, and their own learning requirements (Grossman et al., 2011; Liu and Xie, 2015).

4.3 Analysis of heterogeneity

This section continues the discussion on the relationship between land loss and academic performance of adolescents from rural families from the gender perspective. Table 7 presents the subsample regression analysis of the academic performance of adolescents, factoring in the gender differences of family decision-makers. The analysis shows that land loss has a negative impact on adolescents' academic performance. However, compared to families with male decision makers, the Chinese scores, math scores, class rank, and grade rank of adolescents from families with female decision makers are higher. This is consistent with the literature suggesting that "compared to men, women usually take greater responsibility for ensuring adolescents' education and thus have a strong preference for educational investment" (Moses, 1977; Xu, 2018; Xinrong et al., 2021; L'Roe et al., 2022). Even under the external risks of land expropriation, women continue to prioritize investment in education.

Table 8 shows the results of the sub-sample regression under the gender difference of adolescents. The analysis reveals that the Chinese and math scores, as well as the class and grade rankings

of boys from land-loss families are significantly lower than those of boys from non-land-loss families. The Chinese scores of boys in land-loss families (influence coefficient: -1.1137) are significantly worse compared to the math scores (the influence coefficient is -0.6950). For girls from land-loss families, the Chinese and math scores of the girls are significantly lower than those of girls from non-land-loss families (the influence coefficient was -0.7970), but the class and grade rankings are not significantly different. The regression results also indicate that the coefficient value of the landloss variable is significantly lower for boys than for girls from landloss families, which is consistent with the literature that suggests that girls tend to maintain superior academic performance despite adverse learning environments (Hall, 1978; Epstein, 1998; Wei and Chen, 2018). Therefore, it can be concluded that the negative impact of land loss on academic performance is significantly greater for boys than for girls, and is consistent with previous studies (Jiankun and Guangye, 2019). Hypothesis 3 is therefore confirmed.

4.4 Test for robustness

In the above discussion, Chinese and math scores, as well as class and grade rankings of adolescents, were selected as the comprehensive measurement indicators of their academic performance. Table 9 shows the robustness test result, where the total score for Chinese and mathematics was taken as the dependent variable. Fixed-effects and PSM-DID models were implemented to measure the differences in academic

TABLE 9 Robustness test.

Variables	Tota	l score	
	FE	PSM-DID	
treated	-1.323**	-1.409**	
	(-2.353)	(-2.504)	
gender	-1.646***	-1.596***	
	(-7.980)	(-7.441)	
agep	2.081***	2.099***	
	(28.642)	(27.479)	
agep2	-0.021***	-0.021***	
	(-23.460)	(-22.583)	
edumax	-0.156**	-0.231***	
	(-2.460)	(-3.180)	
fs	0.010***	0.011***	
	(4.532)	(4.758)	
young	0.352**	0.407***	
	(2.334)	(2.583)	
old	1.068***	1.0720***	
	(7.664)	(7.366)	
lnfincome1_per	0.212**	0.453***	
	(2.012)	(2.811)	
lntotal_asset	-0.341***	-0.345***	
	(-3.202)	(-3.058)	
fm	-0.324	-0.229	
	(-0.881)	(-0.607)	
cg	-0.014	-0.008	
	(-1.186)	(-0.682)	
ch	0.007	0.003	
	(0.664)	(0.265)	
provcd	-0.234*	-0.295**	
	(-1.835)	(-2.216)	
Adj. R ²	0.269	0.271	

Note: total score = math score + Chinese score; t-value in parentheses; ****p < 0.01 **p < 0.05 *p < 0.1.

performance between adolescents from land-loss and non-land-loss families. The results show that the coefficient value of the land-loss variable is significantly negative. The impact effect (the absolute value of the coefficient) is significantly larger than the impact of Chinese or math scores (the impact coefficients of Models 3 and Model 4 are -0.7352 and -0.6747, respectively). This indicates that adolescents from land-loss families have significantly lower overall performance than those from non-land-loss families and land expropriation will produce a negative

intergenerational transmission effect. Therefore, these results confirm that the above regression results are robust.

5 Discussion

The rapid pace of urbanization has led to a significant rise in the number of land-loss farmers. These farmers experience passive urbanization, which forcefully separates them from the

traditional agriculture-based rural community, thereby altering their institutional identity (Kumar et al., 2021). Additionally, due to low labor skills, they have an incomplete establishment of the employment relationship network that reflects the modern dimension. Together, the lack of necessary psychological transition and adjustment in the process of urbanization hinders land-loss farmers from adapting to city life. Therefore, it is a realistic problem to consider whether they can integrate into urban life and adjust to urban production immediately after losing their land.

The literature includes a follow-up survey on the income level and life quality of land-loss farmers, revealing that most land-loss farmers experience negative effects during urbanization, such as income decline and a high unemployment rate, which negatively affect the physical and mental development of their children due to changes in their parents' social status (Qin, 2003; Chi, 2004). Adolescence is a critical stage in the formation of individual human capital, and this study focuses on the adolescent group to supplement and verify the relationship between land-loss and adolescent academic performance. The results in Table 3 show a significant negative effect on adolescents' academic performance, confirming that land loss weakens adolescent academic performance and indicating that the negative impact of landlessness on the human capital of family members is intergenerational transmission. This puts the whole family at risk of remaining in chronic poverty, as landless farmers have difficulty entering the higher labor force, and their children lack the ability to access quality educational resources. When farmers' income decreases, they are more likely to prioritize livelihood security over their children's learning and discipline requirements, directing more time and energy toward acquiring employments and constraining their children's learning and educational expenditure, ultimately reflecting a decline in academic performance. This confirms the influencing mechanism of land loss on the academic performance of adolescents based on the logic of "land loss→ family education cognition →family human capital investment→ adolescents' academic performance." Investment in human capital as a mediating mechanism, including both material educational expenditures and immaterial parenting styles, is actually caused by the deterioration of family economic conditions after land loss, and post-loss work status is critical to block or enhance the impact of land loss on children's development (Liu and Xie, 2015; Ma and Lin, 2019). To further alleviate the negative impact of land loss on adolescents' academic performance, vocational training should be strengthened to produce higher marginal benefits and to effectively overcome the unemployment risk.

Previous studies have shown that mothers have a strong preference for educational investments (Wang and Cheng, 2021). Compared to men, women pay more attention to their children's food, clothing, housing, transportation, education, and educational investment, being more motivated to invest in their children's human capital formation (Xinrong et al., 2021). The results of the present study confirm this finding. Children's academic performance is higher in families with female decision makers than in families with male decision makers. Thus, compared to other family members, mothers have a greater influence on children's academic performance, being more inclined to allocate resources toward education.

The boy crisis is a serious issue not limited to China, but observed globally. Boys falling behind in academics will have a significant impact on individuals and society. Moreover, the boy crisis is not limited to academics; boys lag behind in terms of mental health, physical fitness, and social adjustment. This study's conclusion supports the boy crisis theory, specifically shown through the effect of land loss on adolescents' gender. The findings of the current study demonstrate that boys' academic performance is significantly more affected than that of girls, and girls can continue their studies even in an unfavorable learning environment. To promote intergenerational investment in human capital, we should focus on optimizing policies to improve the academic performance of land-loss adolescents and pay more attention to boys. In conclusion, this study will help protect the benefits of land-loss farmers, reduce the risk of unemployment, prevent the adverse impact of land loss on adolescents' human capital investment and academic performance, and help the landloss group escape poverty.

6 Conclusion and suggestions

In order to identify the impact and mechanism of land loss on the academic performance of rural adolescents, this study utilizes data from the China CFPS 2014 and 2018 and applies the PSM-DID and KHB models to explore the effect and mechanism of land loss on rural family human capital investment and adolescents' academic performance based on the cognition-investment-performance framework. The results indicate that land loss has negative effects on rural families' education expenditure and adolescents' academic performance, verifying the logical mechanism of "land loss→ family education cognition→ family human capital investment→ adolescents' academic performance". Therefore, the following conclusions can be drawn:

The academic performance of adolescents from land-loss families was found to be lower than that of adolescents from non-land-loss families, as measured by class and grade ranking, as well as Chinese and mathematics scores.

Land-loss families were found to have lower awareness of educational value compared to non-land-loss families. As a result, they divert their attention from their children's learning and discipline requirements, leading to lower family spending on children's education.

The logical mechanism of "land loss \rightarrow family education cognition \rightarrow family human capital investment \rightarrow adolescents' academic performance" was verified. The differences in the academic performance of adolescents from land-loss families were found to be jointly affected by educational value cognition, educational resource acquisition scale, and their own learning requirements.

The negative effects of land loss on adolescents' academic performance showed gender differences. The academic performance of adolescents in families with female decision makers was found to be higher than that of adolescents in families with male decision makers, and the academic performance of boys in families with land loss was significantly more affected than that of girls.

Based on the conclusions, we suggest the following policy recommendations:

Improving income provides a good economic foundation for adolescents' education to reduce academic performance loss. This can be achieved through targeted and improved professional training and educational programs for land-loss farmers, as well as differentiated employment and vocational skill training for their diverse needs. An employment information platform for land-loss farmers should be established, and relevant preferential policies should be formulated to guide land-loss farmers in starting businesses and helping them accumulate specific human capital.

The social security system should be strengthened by providing suitable forms of old-age security, expanding the coverage of social insurance, and including land-loss farmers in the urban pension and medical insurance system. The unemployment insurance and assistance system should be improved to ease the pressure of reemployment. Moreover, an educational target management file system for adolescents of land-loss families should be established to ensure they complete school and improve their academic performance.

Employment training and job information for women should be provided to optimize their employment environment. A hierarchical assistance strategy should be implemented to support married and child-bearing women who have lost their land. Family support can be strengthened through sharing more housework and childcare responsibilities and enhancing women's family status. Attention should be paid to the educational needs of boys. Adolescents' mental health education should be strengthened to obtain maximum development opportunities and reduce the impact of land loss on boys' learning status.

This study, however, has some limitations. Although it provides empirical evidence from rural areas in China, it is limited by the use of CFPS survey data. Adolescents' academic performance is obtained by asking parents to evaluate their children's performance, which is subjective and may be biased. Furthermore, while this study provides empirical evidence on the relationship between land acquisition and adolescent academic performance in China, it is unclear whether the findings apply to other countries with different national conditions or resource endowments, particularly in countries with severe land loss conditions. Moreover, in 2021, China introduced a "double reduction policy". Whether the impact of land loss on the human capital accumulation of rural adolescents will change as a result of this policy is a topic worth investigating. Therefore, future research should aim to collect more regional data, describe adolescents' academic performance more objectively, and incorporate real social variables to deepen our understanding of the influence mechanism between the two.

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Data availability statement

Publicly available datasets were analyzed in this study. This data can be found here: http://www.isss.pku.edu.cn/cfps/.

Author contributions

Conceptualization, JH; methodology, JH; software, JH; validation, JH; formal analysis, JH and RL; investigation, JH; resources, JH; data curation, JH; writing—original draft preparation, JH and RL; writing—review and editing, JH and RL; visualization, JH and RL; supervision, JH and RL. All authors have read and agreed to the published version of the manuscript.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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