APPLICATION OF HIGH-TECH AGRICULTURE BY HOUSEHOLDS IN THE RED RIVER DELTA OF VIETNAM

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Abstract: This study is conducted to model the factors that influence the decision to apply high technology in agricultural production (agricultural production) of farmers in the Red River Delta (RRD), Vietnam. Data used for the study are surveyed in fact 600 households participating in agricultural production in the RRD. The paper uses the Binary Logistic regression analysis method to identify the factors affecting the decision to apply high technology to agricultural production. The research results show that factors such as farm households' education, people's participation in social organizations, the impact of urbanization, capital and market availability all have positive effects extreme decision to apply technology of farmers. Based on the results of the study, the paper proposed policies to be focused on to enhance the application of high technology in agricultural production by farmers. The findings of this study demonstrate that people's application of high technology to agricultural production is improved by enhancing their production capacity.

Key words: High-tech agriculture, agricultural production, technology application, Red River Delta, Vietnam

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INTRODUCTION

From the new pressures of international economic integration and the impact of climate change, the need for the development of high-tech agriculture (HTA) is put at the forefront towards the goal of improving efficiency, breakthrough in productivity, and quality of products. HTA is an agriculture that applies new technologies to production including agricultural technology (ie production mechanization), automation, biotechnology, information technology, new materials and plants and animals with high productivity and quality, high efficiency per unit acreage and develop sustainably on the basis of ownership, at the same time, HTA can also protect the environment through reducing use of agrochemicals (Xo and Nhuong, 2006; Wolf and Terrell, 2016). The Covid -19 pandemic took place, once again affirming that agriculture is a safe platform for countries with agricultural advantages like Vietnam (Gregorioa and Ancog, 2020). Vietnam is considered an agricultural country with over 80% of the population in rural areas and about 74.6% of the population working in agriculture. In recent years, growth and development of the agriculture in Vietnam, according to some studies, is still mainly based on the use of human resources, natural resources and input materials (Tri, 2021).

At the same time, the value of agricultural products is still low because most of them are still in raw form and their quality is not high, their competitiveness is low, they have not really used effectively and brought into play all resources. The RRD is a large land located around the lower Red River in northern Vietnam. The land of the region is very suitable for intensive wet rice cultivation, crops and short-term industrial crops. The region has the second largest growing area of food crops in the country with an area of 1242.9 thousand hectares. Most of the provinces in the RRD (Ha Nam, Nam Dinh, Thai Binh, Hai Duong, Ninh Binh, etc.) lead the country in land consolidation, land accumulation to form large

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fields, bringing mechanization into production to improve labor productivity, thereby forming concentrated production areas, such as: The carrot growing area in Gia Binh is 700 ha wide and the potato growing area in Que Vo is nearly 1,500 ha (Bac Ninh) ; the carrot growing area in Cam Giang (Hai Duong) is over 500 ha wide, the onion and garlic growing area in Kinh Mon (Hai Duong) is over 3,000 ha; flowers and ornamental plants in Van Giang (Hung Yen) are over 250 hectares wide, etc. (Chinh, 2020). However, the process of applying high technology to agricultural production in the RRD is still slow, inconsistent and spontaneous (Dang et al., 2011; Luu et al., 2019). Therefore, this study selected the RRD region as a typical model for the study. The topic of high-tech application in agricultural production is also quite attractive to interested researchers (Biwas, 2010). Some recent empirical studies show that the level of access to high technology depends on many factors such as demographics, production conditions of farmers, policies, extension services, market (Li et al., 2019; Baffoe-Asare et al., 2013; Son and Thanh, 2014).

Many methods are applied to evaluate the factors affecting the decision to apply industrial technology, such as the Tobit regression method, which is used to evaluate socio-economic factors affecting the decision to apply public works high technology (CODAPEC) of farmers in central Ghana (Baffoe-Asare et al., 2013; Eneji et al., 2012; Thien and Hanh, 2019); structural modeling method (SEM) to determine the factors affecting the degree of agricultural mechanization in China (Li et al., 2019); the regression and correlation analysis method is used to examine the relationship between the factors affecting the application of agricultural technology by smallholder farmers in Kakamega (Kinyangi, 2012); logistic regression method and factor analysis (EFA) are used to evaluate the factors affecting the decision to apply technical advances and the degree of application of technical advances to rice production of farmers in Hau Giang province (Son and Thanh, 2014). In this study, the authors choose the Binary Logistic Model to identify the factors that affect the farmers' decision to apply high technology in agricultural production.

Since the Binary Logistic Model is used to estimate the probability of an event, the feature of the model that the dependent variable has only two values of 0 and 1 is called a binary variable. When the dependent variable is in binary form, it cannot be analyzed with conventional regression because it will violate assumptions, most clearly when the dependent variable has only two expressions, it is not suitable to assume that the remainder has the normal distribution, but instead it will have the binomial distribution, which would statistically invalidate the tests in the normal regression (Son and Thanh, 2014; Loc, 2019; Phuong and Hien, 2014). This study aims to analyze the factors that affect the farmers' decision to apply high technology to agricultural production. The main objectives that need to be addressed in the article are: (i) Building a model of factors that affect decisions on high technology application and determining analytical methods; (ii) Analyze the results from the model, discuss and draw limitations of the research, and at the same time give the next research direction; (iii) Proposing a number of recommendations to promote the agricultural technology development of farmers in the RRD in particular and Vietnamese farmers in general.

MATERIALS AND METHODS 1. Study Area

The RRD region extends from latitude 21°34 N (Lap Thach district) to the alluvial area about 19°5 N (Kim Son district), from 105°17 E (Ba Vi district) to 107°7 E (Cat Ba Island). The RRD including 10 provinces and cities such as Vinh Phuc, Hanoi, Bac Ninh, Ha Nam, Hung Yen, Hai Duong, Hai Phong, Thai Binh, Nam Dinh, Ninh Binh. The whole region has an area of 20.973 km², accounting for 7% of the total area of the country. The population density of the region is the highest in Vietnam (1064 people/km², population is 22 million). The area of agricultural land is about 760,000 ha, of which 70% is fertile alluvial soil, which has great value for agricultural production. Agricultural land occupies 51.2% of the area. Three

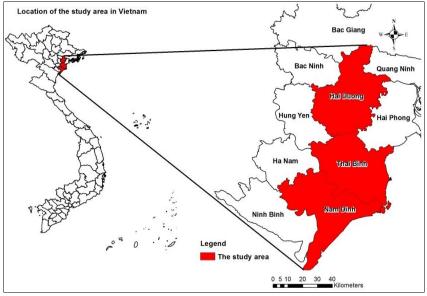


Figure 1. Location of three selected provinces in the RRD, Vietnam

provinces including Nam Dinh, Thai Binh, Hai Duong are selected as study areas because they are the key agricultural areas in the RRD. Nam Dinh is a province located in the south of the Northern Delta (also known as the RRD). Nam Dinh province stretches from 19°54'N to 20°40'N and from 105°55'E to 106°45'E. To the north, it borders Thai Binh province. To the south, it borders Ninh Binh province. The West borders Ha Nam province.

East sea bordering the east. (The administrative atlas of Vietnam, 2013). In 2020, The province's gross domestic product (GRDP) was estimated at 46,121 billion VND, up 5.5% over the previous year. In which, the agriculture, forestry and fishery sector was 9,590 billion VND, up 3.99%. Economic structure: The agriculture, forestry and fishery sector accounts for 22.46%; industry and construction accounted for 39.48%; service sector accounted for 34.93%; product tax minus product subsidies accounted for 3.13% (namdinh.gov.vn).

Thai Binh is a coastal province in the RRD. Thai Binh province has coordinates from 20°18'N to 20°44'N, 106°06'E to 106°39'E. The North borders Hai Duong province, Hung Yen province and Hai Phong city. The West borders Ha Nam province. To the south, it borders Nam Dinh province. East sea bordering the east. (vi.wikipedia.org). Thai Binh is a coastal province in the RRD, with a natural area of 1,534.4 km², a population of about 1.8 million people. Thai Binh is a coastal province in the RRD, with a natural area of 1,534.4 km², a population of about 1.8 million people. 2020, Gross domestic product in the GRDP is estimated at 53,523 billion VND, an increase of over 3.2% compared to 2019; in which the agriculture, forestry and fishery sector increased by 3.4%. The per capita income of Thai Binh province in 2020 is estimated at 50.76 million VND/ person / year (thaibinh.gov.vn). Hai Duong stretches from 20°43' to 21°14' north latitude, 106°03' to 106°38' east longitude. The North borders Bac Giang province. To the south, it borders Thai Binh province and Hai Phong city. The West borders Bac Ninh province and Hung Yen province. To the south, it borders Thai Binh province. Hai Duong has an area of 1,662 km². In 2020, Hai Duong is the 8th largest administrative unit in Vietnam in terms of population with 1,916,774 people, the GRDP growth rate is 2.1%. GRDP reached 134,700 billion VND (equivalent to 5.790 billion USD), GRDP per capita reached 69.8 million VND (equivalent to 3,020 USD) (vi.wikipedia.org).

2. Basis for building research model

Based on the conceptual framework of Kinyangi (2012); Wei et al. (2019); Kinyangi (2012); Bucci et al. (2019); Anh (2012); Baffoe-Asare et al. (2013); Son and Thanh (2014); and in combination with consulting experts and results of the trial investigation, the research selected Binary Logistic model with the following variables: $\ln \left[\frac{p(Y=1)}{p(Y=0)} \right] = \beta_0 + \beta_1 \text{ NOE}$

+ $\beta_2 EDU + \beta_3 PSO + \beta_4 CAP + \beta_5 EXP + \beta_6 URB + \beta_7 MAR$

In which: Y is a binary variable, representing the application of high technology of a farmer to production and is measured by two values 1 and 0 (1 is a farmer with at least one high-tech model application, 0 is farm households that do not apply high technology or are called traditional agricultural households).

 $P(Y = 1) = P_0 \text{ is the probability of applying high technology;}$ $P(Y = 0) = 1 - P_0 \text{ is the probability of not applying high technology;}$

$$Ln \frac{\mathbf{P}\mathbf{0}}{\mathbf{1-P}\mathbf{0}} = \mathbf{\beta}_0 + \beta_1 \text{ NOE} + \beta_2 \text{ EDU} + \beta_3 \text{ PSO} + \beta_4 \text{ CAP} + \beta_5 \text{ EXP} + \beta_6 \text{ URB} + \beta_7 \text{ MAR}$$

Odds: $O_0 = \frac{\mathbf{P}\mathbf{0}}{\mathbf{1-P}\mathbf{0}} = \frac{P(applying \ high \ technology)}{P(not \ applying \ high \ technology)}$ Thus, the Binary Logistic regression function on factors

affecting the decision to apply technology in agricultural production is as follows:

Ln O₀ = $\beta_0 + \beta_1$ NOE + β_2 EDU + β_3 PSO + β_4 CAP + β_5 EXP + β_6 URB + β_7 MAR

Variable		Interpretation	Source	Expectation
	Number of employees (NOE)	Total number of employees directly involved in agricultural production in the household of a farmer.	Li et al., 2019; Son and Thanh, 2014	+
	Education (EDU)	Years of schooling that the head of the household has studied up to the time of the study.	Baffoe - Asare et al., 2013; Bucci et al., 2019	+
	Participation in a social organization (PSO)	1 = farmer participates in a social organization 0 = farmer participates in any organization	Son and Thanh, 2014	+
Independent variables	Capital (CAP)	1 = farmer has a loan 0 = farmer has no loan	Baffoe-Asare et al., 2013; Kinyangi, 2012.	
	Experience (EXP)	Number of years the household head was participated in agricultural production as of the study time	Richard Baffoe - Asare et al., 2013	+
	Urbanization (URB)	Farmers' assessment of the affected urbanization process agricultural production activities (1 = completely unaffected;; 5 = completely affected)	Anh, 2012	+
	Market (MAR)	Farmers' assessment of market availability for agricultural products ($1 = \text{complete}$ do not affect;; $5 = \text{refund full influence}$)	Anh, 2012; Kinyangi, 2012	+
Dependent variable	Decision on application of high technology in agricultural production (DAHT)	1 = applying high technology 0 = not applying high technology	Kinyangi, 2012; Son and Thanh, 2014	

Table 1. Interpretation of the variables in the model (Source: Authors synthesized, 2020)

3. Sampling method

To ensure the science and representativeness of the primary data, the minimum sample size to achieve is calculated by the formula N> 50 + 8 * m (m: number of independent variables) [24]. This study consists of 7 independent variables, so the minimum sample number will be N = 50 + 8 * 7 = 106 survey samples. However, in order to have accurate results, the topic has arranged to survey 600 questionnaires to ensure the number of questionnaires. The data collection includes the steps:

Step 1: Contacting the survey site to select the study area, consulting with local officials and experts to select the site. After consulting, the author selected survey locations in 3 highly representative provinces: Nam Dinh, Thai Binh, Hai Duong.

Step 2: Based on the results of document review, consult experts and local practices, research to build and complete the initial survey. After that, the authors group conducted a trial investigation of 50 questionnaires to check the suitability of the survey, and at the same time correct the survey.

Step 3: After editing the survey questionnaire, the authors conduct an official survey with a sample size of 600 farmer households. The survey process will be conducted from May to July 2020.

After collecting and synthesizing the questionnaires, the study used SPSS 20 software to analyze the data. The specific analysis steps are as follows: Descriptive statistical analysis; Analysis of tests; Binary Logistic regression results discussed.

RESULTS AND DISCUSSION

1. Descriptive statistics

Based on the results collected from the survey, there are 32 unsatisfactory responses, 568 valid votes are collected, the results are presented in Table 2. Out of 568 interviewees, female farmers are more likely to respond than men, this could also indicate that more men are working in factories, companies, and institutions. more organizations than women; Women stay mainly at home, farming and taking care of the children.

The age group to participate in interviews is relatively equal, but the groups that respond the most are 20-29 and 40-49 years old. The results indicate that experience, knowledge, competencies and skills seem to increase with an increase in years at work, in which the group with more than 11 years of experience responded more with 66.72%. Most of the farmers here have education from lower secondary school and above and are not illiterate.

Table 2. Demographic characteristics (Source: According to the author's survey, 2020)

Characteristics	Percent of frequency		
1. Age	N = 568	100%	
20 - 29 years	261	45.95	
30 – 39 years	104	18.31	
40 – 49 years	203	35.74	
2. Sex	N = 568	100%	
Female	389	68.49	
Male	179	31.51	
3. Experience	N = 568	100%	
Less than 5 years	74	13.03	
5-10 years	115	20.25	
11 – 16 years	255	44.89	
Over 16 years	124	21.83	
4. Academic level	N = 568	100%	
Primary school graduated	49	8.62	
Secondary school graduated	308	54.23	
High school graduated	186	32.75	
College and University graduated	25	4.40	

2. Binary Logistic Regression Results

2.1 Analysis of tests

a) Verification of the regression coefficient

Table 3. Results of regression analysis using Binary Logistic

mode (Source: According to the author's survey, 2020)				
В	S.E.	Wald	Sig.	Exp(B)
3.546	1.303	8.347	0.000**	39.116
2.782	0.890	7.552	0.001**	14.174
1.584	0.412	12.612	0.004***	4.511
2.409	1.174	3.604	0.039**	10.801
1.764	0.671	5.723	0.026**	6.281
-48.302	12.832	13.871	0.007	0.000
	B 3.546 2.782 1.584 2.409 1.764	B S.E. 3.546 1.303 2.782 0.890 1.584 0.412 2.409 1.174 1.764 0.671	B S.E. Wald 3.546 1.303 8.347 2.782 0.890 7.552 1.584 0.412 12.612 2.409 1.174 3.604 1.764 0.671 5.723	B S.E. Wald Sig. 3.546 1.303 8.347 0.000** 2.782 0.890 7.552 0.001** 1.584 0.412 12.612 0.004*** 2.409 1.174 3.604 0.039** 1.764 0.671 5.723 0.026**

Note: ***, **, * are significant at 1%, 5%, and 10%

Table 4. Predictive classification (Classification Table) (Source: According to author's calculations, 2020)

Observation		Prediction		
		Decision		Percentage
		No	Yes	Correct (%)
Decision	No	224	12	94.92
Decision	Yes	45	287	86.45
Overall Percentage			92.08	

Table 5. Omnibus Tests of Model Coefficients (Source: According to the author's survey 2020)

(Source: According to the aution's survey, 2020)				
		Chi-square	df	Sig.
	Step	211.385	5	.000
Step 1	Block	211.385	5	.000
_	Model	211.385	5	.000

The regression results with 7 initial independent variables showed that, Sig (EXP) = 0.818 > 0.1 and Sig (NOE) = 0.213 > 0.1, so the two variables of experience (EXP) and number of employs (NOE) are not statistically significant. Statistics or not affect the decision on application of high technology in agricultural production (DAHT). The remaining variables are education, participated in a social organization, capital, urbanization and market, with Sig values of 0.003, 0.016, 0.004, 0.032, 0.000 <0.05 respectively, so it is statistically significant and influential to decide to apply high technology in agricultural production (DAHT). After eliminating two experience variable (EXP) and number of employs variable (NOE), Binary Logistic was performed with SPSS 20, and obtained the following results:

From the results of the above regression analysis, we can see that the significance level Sig of the variables has the value <0.05, so the independent variables in the Binary Logistic regression model correlated with the dependent variable is decision to apply high technology in agricultural production (DAHT).

b) Verification of the suitability of the model

Verify the forecast accuracy of the model: The results in Table 4 show that there are 236 observing cases of not applying high technology in agricultural production, predicting there are 224 cases of not applying high technology in agricultural production rate is correct. 94.92%. There are 332 observational cases with high technology application in agricultural production, predicting there are 287 cases of high technology application in agricultural production rate of 86.45%. Hence, the correct prediction rate of the whole model is 92.08%.

Check the suitability of the model: Based on the results of the Omnibus test on the suitability of the model in Table 5, we have Chi-square = 211.385 and Sig = 0.000 < 0.05, so the independent variables have a linear relationship with dependent variable in the population. The model is suitable.

In Table 6, using the results of the regression coefficient column (B) and column Exp (B), form a scenario where the probability changes when the initial probabilities are 10%, 20%, and 30%, respectively 40% and 50%. The results are as follows:

Variable CAP: Assuming the probability of applying technology of the original farmer is 10%. When other factors do not change, if a household has a loan, the probability of this household's high-tech application will increase to 81.3%.

Similarly, 90.72%, 94.37%, 96.31%, 97.51% respectively when the initial probabilities are 20%, 30%, 40% and 50%. This result shows that the capital variable has a positive impact on the decision to apply high technology in agricultural production, with results consistent with the study (Kinyangi, 2012; Baffoe-Asare et al., 2013; Son and Thanh, 2014). If farmers are supported with loans, they will have the opportunity to invest in modern equipment and machinery in agricultural production. This investment in the long term will help farmers save costs such as labor costs, mechanized hiring costs, etc. thereby helping farmers increase profits in production.

Discussion 2.2. on regression results

Table 6. Simulate probability of applying high technology in
agricultural production (Source: According to the author's survey, 2020)

20

90.722

77.990

53.002

Simulate the probability of applying technology when independent

variable has changed 1-unit and the initial probability is: %

30

94.370

85.864

65.908

40

96.306

90.430

75.045

87.805

80.722

50

97.507

93.409

81.854

91.526

86.26562

Calculate and simulate the applying probability of high technology in agricultural production by the following formula: Poxe^β

$$P_1 = \frac{1}{1 - P_0(1 - e\beta)}$$

Where: P0 is the initial probability;

EDU 1.584 $1 - PO(1 - e^{\beta})$ URB 2.409 MAR

82.234 10.80 54.547 72.974 1.764 6.28 41.103 61.093 72.913 P1 is the change probability Variable PSO: Assuming the probability of applying technology of the household is initially 10%. When other factors

Exp

(B)

39.12

14.17

4.51

10

81.295

61.163

33.387

В

3.546

2.782

Variable

CAP

PSO

do not change, if a household participates in social organization, the probability of applying high technology of this household will increase to 61.16%, Similarly, 77.99%, 85.86%, 90.43 respectively. %, 93.41% when the initial probabilities are 20%, 30%, 40% and 50%. This result shows that the participation in social organization variable has a positive impact on the household head's decision to apply high technology in agricultural production, with consistent results (Son and Thanh, 2014; Kinyangi, 2012). That's right, because through social organizations, farmers are exposed to different strata in society, helping people to expand their knowledge, helping them realize the progress of science. reporting by participating in training courses, propaganda and seminars of social organizations. From there, it helps people to change perceptions and give up outdated farming habits.

Variable EDU: Assuming that the probability of applying technology of the household is 10% initially. When other factors do not change, if the number of years of schooling of the household head increases by one unit, the probability of applying high technology of this household will increase to 33.39%. Similarly, 53%, 65.91%, 75.05%, 81.85% respectively when the initial probabilities are 20%, 30%, 40% and 50%. This result shows that the education variable has a positive impact on the household head's decision to apply high technology in agricultural production, with results corresponding to the study (Kinyangi, 2012; Li et al., 2019; Baffoe-Asare et al., 2013). This is very true in fact, when the education level of the household head is higher, the ability to grasp high-tech equipment is higher than that of those with low education. Therefore, it has increased the ability to decide on the application of high technology in agricultural production.

Variable URB: Assuming the probability of applying technology of the original farmer is 10%. When other factors do not change, if the impact of urbanization increases by one unit, the probability of high technology application of this household will increase to 54.55%, Similarly, 72.97%, 82.24%, 87.81%, 91.23% respectively when the initial probabilities are 20%, 30%, 40%, and 50%. This result shows that urbanization variable has a positive impact on the decision to apply high technology in agricultural production. Urbanization has a significant impact on agricultural production, such as reducing arable land area, and reducing productivity. This motivates farmers to apply high technology to production to increase productivity and quality to meet market demand. Not only that, but urbanization also helps farmers to easily absorb technological advances, helping people no longer fall out of date in production (Anh, 2012).

Variable MAR: Assuming that the probability of applying technology of the household is initially 10%. When other factors do not change, if availability of market increases by one unit, the probability of this household's high technology application will increase to 41.1%. Similarly, 61.09%, 72.91%, 80.72%, 86.27% respectively when the initial probabilities are 20%, 30%, 40%, and 50%. This result shows that the availability of market variable positively influences the decision to apply high technology in agricultural production, this result corresponds to the study (Kinyangi, 2012). When markets are available, agricultural products are consumed quickly, and farmers have more incentive to produce. Because if there is no market for consumption, the leftover goods cause great damage to farmers because most agricultural products have a very short shelf life. Thus, revenue will decrease and farmers do not have money to pay for and invest in the next crop, so the application of technology will also decrease.

CONCLUSION

The research has built up a model of factors affecting the decision to apply high technology in agricultural production of farmers in the RRD. Based on the results of the Binary Logistic regression analysis and tests, it can be confirmed that the factors have affected the high technology application of the farmers include the farmer's education, participation in the social organization, the impact of urbanization, capital and availability of market is positive. The results of this study show that policies should be focused on enhancing the application of high technology, including establish appropriate policies to encourage households to borrow capital to invest in agricultural production. At the same time, the authorities strengthened the direction of the banking system to step in so that farmers could easily and conveniently access loans.

There are appropriate plans to quickly implement the work of *land consolidation and exchange*, on that side, do well the planning of fields, irrigation, and intra-field traffic to meet the requirements of synchronous mechanization.

Increase funding for training for technical staff, households, farm owners, agricultural enterprises on skills in operating and maintaining agricultural machinery and equipment. Implementation of agricultural planning, regional planning and regional linkage in the direction of promoting local comparative advantages, and based on domestic and international market demand, should not be fixed area of rice, so conservation of agricultural land.

Enhance the role of social organizations such as the Farmers' Union, Women's Union, Youth Union and the Vietnam Consumers Association in encouraging the application of high technology in agricultural production and replication of examples, as well as promoting the link between farmers and enterprises in the chain of production, breeding and consumption of products. Although the initial research objectives have been set, the research still has certain limitations. Firstly, the survey was only conducted in 3 provinces of Nam Dinh, Thai Binh and Hai Duong, so the scope of research is limited. Secondly, the number of respondents who are female respondents is higher than that of men, gender equality among the respondents to this study is not achieved, so the representative of respondents may be limited. The next research direction for these studies is to compare the differences in the efficiency of agricultural production by the households using high technology and the group that does not use high technology in agricultural production.

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