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Technical efficiency among agricultural households and determinants of food security in East Java, Indonesia

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Rice is a staple food in East Java, and the average consumption is 100 kg/capita/year. However, rice productivity has declined dramatically in recent years. Food security can be reached by improving the technical efficiency of rice farming, especially in rice farming centers such as East Java Province. This study aims to measure technical efficiency and its determinants using two limit tobit. And it also aims to examine the effect of the technical efficiency of rice farming on food security using logit regression. Technical efficiency will be measured by using data envelopment analysis (DEA). The results show that the technical efficiency of rice farming is very low in East Java. Government assistance, irrigation, and extension have a significant effect on technical efficiency. Meanwhile, membership of farmer organization has no effect on technical efficiency. Around 69% of farmers can be categorized as food secure households. The estimation of logit regression shows that household size, income, land size, education, age, and gender significantly influence food security in East Java. Meanwhile, credit and technical efficiency did not have any significant effect.

Agriculture is a dominant sector in Indonesia, including in East Java Province. East Java Province was the largest producer of rice during 2013–2017¹. The lowest layer of Indonesia's population pyramid is farmers and fishermen who live in rural areas. There are a lot of people working in the agricultural sector. The data shows that the population of East Java aged 15 years and over amounted 6,054,066 people are more likely to work in the agricultural sector (crops, horticulture, plantations, and livestock) which are divided into 3,591,231 men and 2,462,835 women. Meanwhile, 13,519,716 people work in the non-agricultural sector. East Java is the province with the largest population working in the agricultural sector compared to other provinces in Indonesia². It indicates the importance of the agricultural sector in East Java.

The agriculture sector is divided into food crops, horticulture, plantation, livestock, forestry, and fisheries sub-sectors. However, this research focuses on investigating food crops. Food crops are very identical to rice/rice plants because most of the population of Indonesia (including East Java) consume rice as a staple food. It is supported by the data from the Ministry of Agriculture which shows that the average consumption of rice is 100 kg/capita/year. Meanwhile, the consumption of tubers, beans, fruit, or meat per capita is less than 10 kg/year¹.

A large number of rice needs should be followed by the production of rice plants that meet these needs. However, the productivity of rice plants must be increased so that the needs of rice plants or food security are maintained.

Figure 1 shows that rice crop productivity in East Java has declined in the last five years (2014–2018), especially since 2016. This situation caused East Java Province had to import 96.51 million USD worth of rice from Thailand had in 2018³. Declining rice crop productivity can be a signal to take corrective steps so that rice production can meet the needs of the people of East Java. Technical efficiency has an important role in increasing rice productivity.

Technical efficiency is influenced by several factors. Haryanto et al. examined the effect of several factors on the efficiency of rice farmers in several regions in Indonesia, including the factor of government assistance and irrigation. The results show that the government assistance has an effect on the efficiency of rice farming in all areas studied even in Indonesia as a whole⁴.

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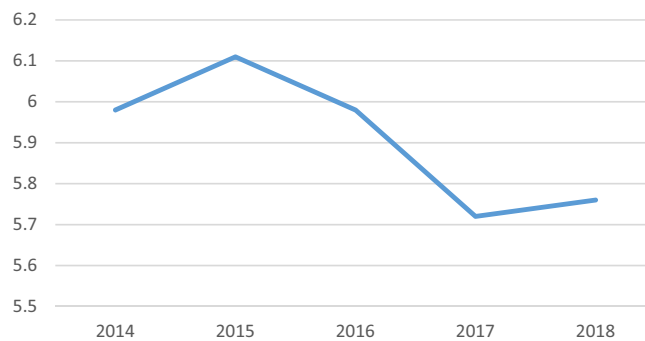


Figure 1. East Java rice crop productivity in 2014–2018 (Ton/Ha). Source: Central Bureau of Statistics.

Karunarithna and Wilson stated that inefficiency is strongly influenced by membership in farmer organizations. Membership in farmer organizations is important because farmers can obtain various facilities such as training, cheap fertilizers, and others⁵. Islam et al.⁶ and San et al.⁷ also stated that extension is very important to increase technical efficiency.

The agricultural sector is certainly much more related to food security compared to other sectors⁸. Several studies related to the efficiency of agricultural production and food security have been conducted several times. Ogundari in his research suggested that efforts to strengthen food security and poverty alleviation can be done by increasing the efficiency of agriculture and food production⁹. Koirala et al. also stated the same thing¹⁰. Food security is a major problem in the Philippines. In addition, food security is greatly influenced by farmers' production decisions, land reform programs, and technical inefficiencies in rice production.

Adeniyi and Dinbabo examined the relationship between technical efficiency and food security using the multinomial regression method. Technical efficiency with SFA showed that the average efficiency was 0.85. It also found that there is a positive and significant relationship between technical efficiency, income and food security¹¹. Majumder et al. in their research stated that there is a way to reduce post-harvest losses of rice and increasing food security in Bangladesh. Improving the technical efficiency of the rice production system in Bangladesh can improve food security¹². Iheke and Onyendi measured technical efficiency and food security using the SFA method and food security index based on daily food intake in Nigeria. The conclusion from this study is: food insecurity is prevalent among agricultural households and food security is very difficult to achieve by farm households even if it is efficient in agricultural management¹³.

Oyakhilomen et al. examined the relationship between technical efficiency and food security of chicken egg-farmers in Nigeria. The method used is SFA, a measure of food security based on the US Department of Agriculture (USDA) and correlation analysis. The results showed that the level of technical efficiency was very low, 23% and 90% of farmers were categorized as food vulnerable. Based on the correlation analysis, it was found that there was a positive relationship between technical efficiency and food security¹⁴. Oyetunde-Usman and Olagunju examined the relationship between technical efficiency and household food security of farmers using the SFA method and the probit model. The results showed that food-resistant farming households were more technically efficient than food-vulnerable households¹⁵. Basically, food security is not only influenced by technical efficiency. There are many more factors that affect food security.

In addition to the efficiency of rice production, there are still many other factors that affect food security. The research conducted by Abdullah et al. revealed that gender has a significant effect on food security¹⁶. Female household heads tend to be vulnerable in maintaining family food security. Agriculture is a sector that is very identical to men's work because it drains a lot of energy and thoughts so it is not suitable for women. Holden and Ghebru in their research stated that climate change significantly affects agricultural productivity so that it will ultimately have an impact on food security¹⁷.

Mishra et al. argued that increasing rice crop production and identifying the causes of inefficiency in rice crop production can help farmers in obtaining higher incomes, increasing food security, and alleviating poverty. Furthermore, research is needed to find out what factors influence inefficiencies in rice crop production so that food security will be maintained¹⁸.

Emran and Shilpi in their study revealed that labor wages affect poverty alleviation and food security in rural areas¹⁹. Mellor and Malik conducted a study in Pakistan and found that the growth of the agricultural sector had a dominant effect on poverty alleviation and food security²⁰. Montaud et al. conducted a study related to agricultural yields and poverty in Nigeria using the Computable General Equilibrium (CGE) method. The results of his research suggested that a long-term decline in agricultural output in Nigeria results in poverty²¹.

Several factors are considered to affect household food security including household assets²²; home ownership²³; household savings²⁴; financial limitations²⁵; education²⁶; livestock ownership²⁷; unemployment and income²⁸; knowledge about food storage, processing, and nutrition²⁹; corruption, fiscal errors, large debts, and inconsistent government policies³⁰; off-farm job³¹; gender³²; family size, land area, soil fertility, access to irrigation, fertilizer use, seed utilization³³; delivery and access to market information, age³⁴; dependency ratio, electricity connection, irrigation availability³⁵; monthly income, family structure³⁶; and the existence of infrastructure³⁷. There are many factors that affect food security. However, each study adjusts to the availability of data in the field when the research was conducted.

Therefore, to improve the structure of the national economy, it is necessary to make improvements in the agricultural sector. The agricultural sector is closely related to food security issues. The efficiency of rice production is certainly expected to be a solution in the process of increasing food security in East Java, Indonesia. This is the basis of interest to examine the relationship between technical efficiency and food security in East Java, Indonesia.

Material and methods

Area descriptions. The data used in this research is secondary data obtained from the Central Bureau of Statistics in the form of Agriculture Survey (Agriculture Household Income Survey) data for the East Java Province in 2013. The samples were 8603 farm households.

Data envelopment analysis. To measure the efficiency of rice production, the Data Envelopment Analysis (DEA) method was used. Data Envelopment Analysis (DEA) is a method of optimizing a mathematical program that measures the technical efficiency of a Decision Making Unit (DMU) and compares it relative to other DMU that use the same type of input and output. DEA formulates DMU as a fractional linear program to find a solution if the model is transformed into a linear program with the weights of input and output. The relative efficiency of DMU in DEA is also defined as the ratio of total weighted output divided by total weighted output (total weighted output).

DEA assumes that each DMU will have a weight that maximizes its efficiency ratio (maximizing total weighted output/total weighted input). The assumption of maximizing the efficiency ratio uses output orientation in calculating technical efficiency. Another orientation is to minimize input, but both assumptions will get the same results.

A DMU is relatively efficient if the dual value is equal to 1 (100 percent efficiency). If the dual value is less than 1, the DMU is considered to be relatively inefficient. The DEA model is divided into two, namely Constant Return to Scale (CRS) and Variable Return to Scale (VRS). There are two orientations commonly used in the efficiency measurement method using DEA, input-oriented and output-oriented.

The efficiency of rice production is analyzed using the Output-Oriented Data Envelopment Analysis (DEA) method with the assumption of Variable Return to Scale (VRS). This efficiency will be estimated by DEAP 2.1.

$$\begin{aligned} & \text{Max}_{\theta_i, \lambda_i} \theta_i \\ & \text{s.t. } \theta_i y_i - Y \lambda_i \leq 0 \\ & X \lambda_i - X_i \leq 0 \\ & \sum \lambda_i = 1 \\ & \lambda_i \geq 0 \end{aligned}$$

The output variable used is rice (IDR). These inputs consist of land area/harvest area (m²), seeds (IDR), fertilizer (IDR), and labor (IDR).

Determinants of technical efficiency. Tobit regression is very adequate to represent the model of the efficiency effect. Two limit Tobit will be used because the value of rice production efficiency is in the range of 0 to 1. The general specification of the tobit model is as follows.

$$\begin{aligned} TE_i^* &= \beta' Z_i + \varepsilon_i \\ TE_i &= L_{1i} \quad \text{if } TE_i^* \leq L_{1i} \\ &= TE_i^* \quad \text{if } L_{1i} < TE_i^* < L_{2i} \\ &= L_{2i} \quad \text{if } TE_i^* \geq L_{2i} \end{aligned}$$

where TE_i^* is a latent variable that represents an index of technical efficiency, TE_i is the dependent variable being reviewed, Z_i is an explanatory variable vector that represents the characteristics of agriculture, β is a vector of parameters to be estimated, ε_i is error term, L_{1i} and L_{2i} are the lower and upper limit. For more details with the following specifications:

$$TE_i = \delta_0 + \delta_1 Z_{1i} + \delta_2 Z_{2i} + \delta_3 Z_{3i} + \delta_4 Z_{4i} + \varepsilon_i$$

where TE_i is the technical efficiency of farmers in rice production; Z_{1i} is government assistance (D = 1 if farmers have accepted the government assistance, D = 0 otherwise), Z_{2i} is dummy membership of a farmer organization (D = 1 if farmer is a member of farmer organization, D = 0 otherwise), Z_{3i} is a dummy for irrigation (D = 1 if irrigated rice fields, D = 0 otherwise), and Z_{4i} is agricultural extension (D = 1 if farmers have accepted agricultural extension, D = 0 otherwise). δ_0 is the parameter to be estimated, ε_i is a random variable that is assumed to be normally distributed. This model will be estimated by the maximum likelihood method with STATA.13.

Food security. There are many definitions of food security. This term usually implies indirectly that people have the same and sustainable economic and physical access to adequate amounts of food nutrition to meet daily calorie needs and to maintain an active and healthy lifestyle. Complex definitions make food security is quite difficult to calculate. Food availability is the basis for food access and food security³⁸, especially at the household level. One way to measure food security can use a simple method of measuring the ratio of household food availability as an indicator of food security based on the opinion of Frelat et al³⁹.

Ridaura et al. measure household food security by looking at potential food availability⁴⁰. This indicator of measurement of potential food availability (PFA) is measured based on the calorie of calories per farm household

	DMU	%
0.000–0.100	457	5.31
0.101–0.200	2735	31.79
0.201–0.300	2745	31.91
0.301–0.400	1486	17.27
0.401–0.500	632	7.35
0.501–0.600	264	3.07
0.601–0.700	110	1.28
0.701–0.800	62	0.72
0.801–0.900	37	0.43
0.901–1.000	75	0.87
Total	8603	100

Table 1. Distribution of technical efficiency.

based on reports on production and consumption of agricultural products and supplies. Dithmer and Abdulai also measure food security based on food energy consumption as indicated by kilocalories (kcal) per day⁴¹. Muraoka says that the amount of food consumption can be used to describe food security⁴². Swindale and Bilinsky measure food security using the Household Dietary Diversity Score (HDDS) method⁴³. To put it simply, farm households that have food security = 1. Whereas farm households that experience food vulnerability = 0.

To determine whether a farmer's household is food resistant or not, the following formula will be used in this research:

$$\text{Food Security Score} = \frac{A + B + C + D}{4}$$

where A = 1 if have a food supply, otherwise A = 0; B = 1 if have enough food supply, otherwise B = 0, C = 1 if never have experienced food shortages, otherwise C = 0; D = 1 if there are no toddlers who have below normal body weight, otherwise D = 0.

Food security scores will range from 0–1. If the food security score = 1 then the farm household can be said to be food security (D = 1). If the food security score is < 1 then the farm household can be said to be food insecure/vulnerable to food (D = 0).

Determinants of food security. The estimation of the food security model will use logit regression because the value of food security is between 0 and 1. The specifications of the food security model will be estimated as follows.

$$L_i = \ln\left(\frac{P_i}{1 - P_i}\right) = \alpha_0 + \alpha_1 Q_{1i} + \alpha_2 Q_{2i} + \alpha_3 Q_{3i} + \alpha_4 Q_{4i} + \alpha_5 Q_{5i} + \alpha_6 Q_{6i} + \alpha_7 Q_{7i} + \alpha_8 Q_{8i} + \varepsilon_i$$

where $\ln\left(\frac{P_i}{1 - P_i}\right)$ is odds ratio for the occurrence of value 1 (food security), Q are factors that explain food security. The factors are Q_1 = household size (in person), Q_2 = income (in log), Q_3 = land size (in log), Q_4 = credit (D = 1 for farmers who have access credit, otherwise D = 0), Q_5 = education (D = 1 for high school education and tertiary education level, otherwise D = 0), Q_6 = age (in year), Q_7 = gender (D = 1 for male, D = 0 for female) and Q_8 = technical efficiency (0–1). $\alpha_1, \alpha_2, \dots, \alpha_7$ are the parameters to be estimated, ε_i is a random variable that is assumed to be normally distributed. This model will be estimated using the Maximum Likelihood Estimation (MLE) method with STATA.13.

Results and discussion

Data Envelopment Analysis estimation results show that the technical efficiency of rice farming is very low in East Java, which is in average of 0.27.

Table 1 shows that most farmers have technical efficiency in the range of 0.101–0.200 and 0.201–0.300 (2735 and 2745 farmer households). Meanwhile, farmers who have very little technical efficiency are in the range of 0.701–0.800, 0.801–0.900, and 0.901–1000 (62, 37, and 75 farmer households). Many things need to be done to improve this technical efficiency. This should receive more attention from the East Java Government because basically technical efficiency is one of the keys to increase rice productivity.

Returns to scale distribution of rice farming in East Java is presented in Table 2. The efficiency scale of most rice farmers in East Java shows decreasing returns to scale (DRS), amounting to 6323 farmer households. Meanwhile, there are increasing returns to scale (IRS) of 1,847 farmer households and constant returns to scale (CRS) of 433 farmer households. It can be concluded that most of the efficiency scale of farmer households obtains the decreasing returns to scale (DRS), which is around 73.50%. Meanwhile, the IRS is around 21.47% and the CRS is around 5.03%.

Table 3 shows the comparison between food secure and non-food secure households. Farmers with food secure category around 69% or 5916 farmers, 31% or 2,687 farmers have a risk of food vulnerability (non-food

	DMU	%
CRS	433	5.03
DRS	6323	73.50
IRS	1847	21.47
Total	8603	100

Table 2. Distribution of returns to scale.

Variable	Food secure				Non-food secure				Mean difference
	n = 5916				n = 2687				
	Min	Max	Mean	STD	Min	Max	Mean	STD	
Household size	1	13	3.73	1.56	1	16	3.78	1.62	0.05
Income	1	902,852	6500.81	17,045.73	3	140,003	3371.47	5205.69	3129.34
Land size	50	250,000	5416.25	8890.38	50	90,800	3248.08	4495.24	2168.17
Credit	0	1	0.09	0.28	0	1	0.07	0.25	0.02
Education	0	1	0.12	0.32	0	1	0.07	0.25	0.05
Age	16	98	52.59	12.58	20	97	51.16	12.72	1.43
Gender	0	1	0.89	0.32	0	1	0.85	0.35	0.04
Technical efficiency	0.017	1	0.27	0.15	0.017	1	0.26	0.14	0.01
Gov. assistance	0	1	0.40	0.49	0	1	0.43	0.50	0.03
Membership	0	1	0.42	0.49	0	1	0.30	0.46	0.12
Irrigation	0	1	0.53	0.50	0	1	0.37	0.48	0.16
Extension	0	1	0.26	0.44	0	1	0.16	0.37	0.10

Table 3. Descriptive statistics.

Variable	Coefficient	Stand. error	t	P
Constant	0.26	0.00	96.13	0.00***
Gov. assistance	-0.00	0.00	-1.95	0.05**
Membership of farmer organization	0.01	0.00	0.45	0.66
Irrigation	0.01	0.00	3.86	0.00***
Extension	0.01	0.00	2.94	0.00***
Log likelihood	4078.65			
LR test	36.23***			

Table 4. Estimation for Tobit model. ***, **, * indicate significance at 1%, 5% and 10% level.

secure). Household size of rice farmers ranges from 1–16 people and the mean difference between food secure and non-food secure households is 0.05. The mean income is 6,500.81 thousand rupiahs for food secure households and 3371.47 thousand rupiahs for non-food secure households. Land size also has a very striking difference, the mean difference is 2168.17 m². Farmers have very minimal access to credit, only 8% of farmers can access credit. Only 12% of the food secure households are highly educated, while only 7% of the non-food secure households are highly educated. The average age of farmers reaches 52 years which can be indicated that most farmers in East Java are categorized as experienced farmers. Gender of farmers is dominated by men, which is around 88%. This is because the agricultural sector is indeed very identical to men's work. Meanwhile, female farmers are only 12%. About 40% of farmers receive government assistance. 42% of food secure households and 30% of non-food secure households join the farmer organization. 53% of food secure households and 37% of non-food secure households get rice field irrigation. 26% of food secure households and 16% of non-food secure households get agricultural extension.

The estimation result for tobit regression is presented in Table 4. Government assistance, irrigation, and extension have a significant effect on technical efficiency. Meanwhile, membership of farmer organization has no effect on technical efficiency.

Government assistance has a negative and significant effect. If farmers receive government assistance, it means that technical efficiency will decrease. This can be due to farmers' lack of knowledge due to low levels of education. Irrigation has a positive and significant effect. Irrigation is very important in rice farming. If the rice fields are adequately irrigated, then the rice can develop well. Extension has a positive and significant effect.

Variable	Coefficient	Odds ratio	Std. error	Z	P value
Constant	-3.34	0.04	0.01	-14.17	0.00***
Household size	-0.03	0.97	0.01	-2.05	0.04**
Income	0.55	1.74	0.13	7.57	0.00***
Land size	0.46	1.59	0.14	5.22	0.00***
Credit	0.09	1.09	0.10	0.97	0.33
Education	0.49	1.64	0.15	5.37	0.00***
Age	0.01	1.01	0.00	4.96	0.00***
Gender	0.18	1.19	0.09	2.44	0.01**
Technical efficiency	0.24	1.27	0.21	1.41	0.16
Log likelihood					-5104.16
LR test					475.82***

Table 5. Estimation for Logit regression. ***, **, * indicate significance at 1%, 5% and 10% level.

The estimation result for logit regression is presented in Table 5. The estimation shows that most of the variables have a significant effect on food security in East Java, except for credit variables which have no significant effect.

Household size has a negative and significant effect with odds ratio = 0.97 and $p = 0.04$. This means that more members in the family will increase the likelihood that the family experience food vulnerability. Income has a positive and significant effect with odds ratio = 1.74 and $p = 0.00$. If the income of farmers increases, it will increase their food security. Land size has a positive and significant effect with odds ratio = 1.59 and $p = 0.00$. It means that if the size of land owned is larger, farmers will increase food security. Education is one of the most influential variables on food security in East Java. Education has a positive and significant effect with odds ratio = 1.64 and $p = 0.00$. It means that if the farmers have a higher level of education, the farmers will increase food security. Age has a positive and significant effect with odds ratio = 1.01 and $p = 0.00$. It means that farmers who have more experiences (indicated by age) will increase their food security. Gender has a positive and significant effect with odds ratio = 1.19 and $p = 0.01$. It means that male farmers will increase the likelihood that the family experience food security. Meanwhile, access to credit has no significant effect on food security.

The result of research on the effect of technical efficiency is quite surprising. Technical efficiency has no significant effect on food security in East Java. The results of this study are different from the results of research by Ogundari and Koirala et al. which stated that the technical efficiency of the agricultural sector influenced food security. The results of this study are also different from those of Ogundari⁹, Koirala et al.¹⁰, Adeniyi and Dinbabo¹¹, Majumder et al.¹², Oyakhilomen et al.¹³, as well as Oyetunde-Usman and Olagunju¹⁴ which stated that the technical efficiency of the agricultural sector affected food security. Meanwhile, the results of research by Iheke and Onyendi are the same with the results of this study which revealed that technical efficiency does not have a significant effect on food security¹⁵.

Data shows that the level of education of farmers in East Java is relatively low. Farmers who passed primary education were 5,414,605 people (84.34%), secondary education were 575,974 people (14.11%), and higher education 63,487 people (1.51%). This is quite alarming because education has a significant effect on food security in East Java. Based on age, farmers aged 15–24 years were 366,145 people (6.05%), 25–59 years were 4,068,510 people (67.20%), and > 60 years were 1,619,411 people (26.75%). It indicates that most farmers have reached a mature age. But the number of farmers aged 15–24 years is enough to attract attention because it is a sign that only a few young people want to work in the agricultural sector in East Java. Problems with education and age should be resolved if the East Java Provincial Government is able to convince the community that agriculture is a viable and profitable source of livelihood. So that people who are young and highly educated will interest to work in the agricultural sector. If this cannot be done, it is not surprising that more people choose to work in the non-agricultural sector.

The low level of technical efficiency of rice farming can be caused by a lack of community knowledge to manage agriculture well. One solution from the government is an agricultural extension. Growth in the number of agricultural extension workers in East Java in 2019 increased by 6.58% compared to the previous period. This is one of the East Java Provincial Government's commitments to increase public knowledge about a good agricultural system so that farmers' hope to become more technically efficient and ultimately increase food security. The fact is that technical efficiency does not have a significant effect on food security. This can be due to very low technical efficiency. In addition, agriculture is no longer the main occupation for some farmers. Farmers usually have other income as the main source to support the household. So it is not surprising that income is more influential than technical efficiency.

Figure 2 shows that rice farmers are divided into several categories: (A) main income from rice farming activities; (B) main income from rice farming activities but have additional income; (C) main income from other activities instead of rice farming activities.

Based on the research sample, there are 63.55% of farmers who have their main income from rice farming activities. 7.36% of farmers have their main income from rice farming activities but still have additional income. In addition, 29.09% of farmers have their main income from other activities instead of rice farming activities.

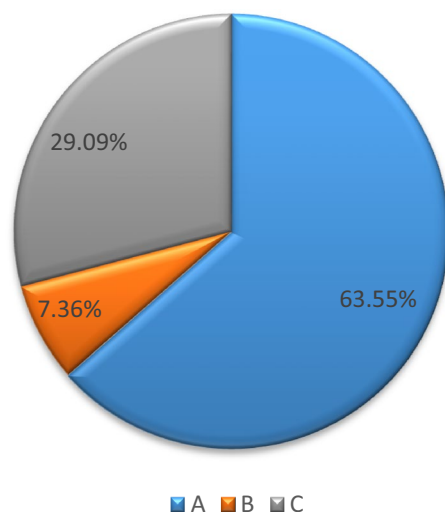


Figure 2. East Java farmer category.

Conclusion

The technical efficiency of rice farming is very low in East Java. Even though East Java is one of the centers of rice farming, Government assistance, irrigation, and extension have a significant effect on technical efficiency. Meanwhile, membership of farmer organization has no effect on technical efficiency.

About 69% of farmers are included in the food secure category, while 31% of farmers are at risk of food vulnerability. The estimation of logit regression shows that household size, income, land size, education, age, and gender significantly influence food security in East Java. Meanwhile, credit and technical efficiency did not have a significant effect. Technical efficiency does not have a significant effect on food security, the government must continue to strive for improving this technical efficiency. The income generated in the agricultural sector also affects food security. The income is related or determined by the productivity of rice farming. The level of productivity must be maintained because if rice productivity decreases, it can certainly reduce the income and food security of farmers in East Java.

Farmer's education is relatively low in East Java. Educated farmers are an important factor because they are expected to be more open-minded about innovations in agriculture to increase technical efficiency. Increasing the efficiency of rice farming can be done by increasing the interest of the educated population to take part in the rice farming management process due to the fact that most of the rice farmers in East Java are still low educated, with the average age is over 50 years old. Most of the farmers are 50 years old and only a small portion of 15–24 years old people work as farmers. In addition, agriculture is no longer the main occupation. Farmers usually have other income as the main source to support the household.

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Competing interests

The authors declare no competing interests.

Additional information

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