Pattern, Trend and Determinants of Crop Diversification: Empirical Evidence from Smallholders in Eastern Ethiopia

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Abstract
Crop diversification is the most important risk management strategies. The study investigated the pattern, trend and covariates of crop diversification in eastern Ethiopia based on data collected from 167 households randomly and proportionately selected. In order to manage risks of drought, pests and diseases, soil fertility decline and input prices variations, farmers in the study areas employ crop diversification as a self-insuring strategy. The farmers are becoming risk-averse which has implications on technology adoption. Tobit model result indicated that farmers with more extension contacts and larger livestock size are likely to specialize whereas those who have access to market information and irrigation, those who own machinery and more number of farm plots are more likely to diversify. In order to promote crop diversification, providing farm machinery through easy loans and improving access to market information and irrigation should be given attention. The extension system should include risk-minimization as a strategy.

Keywords: crop diversification, risk, risk management strategies, risk-averse, Ethiopia.

1. Introduction
Ethiopia is a country whose economy is significantly dependent on agriculture sector. The sector is dominated by smallholder peasant agriculture which accounts for more than 95% of the total food production. The sector supports employment for 85% of the population, 90% of exports and 50% of GDP.
Out of the total agricultural land, about 97% is cultivated by smallholders who produce more than 97% of the agricultural output. Despite its importance, the agricultural sector is traditional and subsistence-oriented (Jema 2008) and it characterized by poor and declining performance. This performance is attributed to many interrelated factors which include recurrent drought, land degradation, paucity of land, crop and livestock pests and diseases, lack of improved and suitable technologies, poor marketing and service infrastructure (Girma 2002; Belaineh 2003).

In Ethiopia, farming is particularly weather-dependent and farmers face price, yield and resource risks that arise from the biophysical and socio-economic environment in which they operate. Rural households in the country are exposed to a variety of risks that include harvest failure as a result of drought, frost, floods and other climatic events and death of livestock (Dercon 2002). These risks influence the production and resource allocation decisions of smallholder farmers. Lack of institutional innovations such as credit and insurance schemes in most developing countries makes individual risk management a significant issue to cash-constrained smallholder farmers.

In an effort to adapt to the risky environment, smallholders make certain production decisions and employ various risk mitigating strategies. Crop diversification is one method of reducing income variability (Briglauer 2000). According to the comparative advantage theory, diversification can reduce risks but at the expense of income. Despite the significant role crop diversification plays in agriculture, there are only few empirical studies on the factors that determine diversification. This study has examined the effect of various household characteristics and other factors on crop diversification.

In Eastern Ethiopia, the economy is characterized by subsistence mixed farming systems, low agricultural productivity and poor access to major markets (Bezabih & Hadera 2006; Jema 2008). Agricultural production in the area is generally characterized by a high degree of instability. This is mainly attributed to the nature of the agricultural production which is associated with high degree of uncertainty which arises from dependency of the agricultural production on uncontrollable weather conditions (erratic and variable rainfall) which caused great fluctuations on crops yield on the one hand, and large fluctuations in input and output prices. It is argued that diversifying by growing more enterprises may lead to farm income stability (Tefera et al. 2003). The smallholder farmers in the area practice crop diversification of different types. However, the pattern and trend of crop diversification and its implications are not studied. The study therefore, attempted to analyze the pattern and trend of crop diversification in eastern Ethiopia.

2. Research Methodology

2.1. The Study Area

The study is conducted in Eastern Hararghe zone which is among the seventeen zones of Oromia National Regional state of Ethiopia. It is located between 7° 32′N-9° 44′N latitude and 41° 12′E-42° 53′E longitude. The Zone has an average altitude of 1750 meters above sea level. Climatically, the zone is subdivided into Kola (note 1), Woinadega and Dega. The zone has a total area of about 24,900 km² and divided into 17 districts. From the zone, Kombolcha and Haramaya districts are selected for this study. Haramaya district is found in the North Central part of the Zone. The district comprises of 33 peasant associations (PAs) and it is characterized by plateau, mountains, hills and plains. Altitudinally stretching between 1400 and 2340 meters above sea level, Haramaya falls under middle high land agro-climatic zone and the mean annual rain fall is 790 mm. The district has a total population of about 236 600 of which about 90% live in rural areas. Kombolcha district is found in the northern part of East Hararghe zone. Physiographically, it is characterized by dissected plateaus, mountains, hills, plains and valleys. Of the 19 peasant associations in the District, seven (37%) are located in the lowlands and the remaining 12 (63%) are located in the Woinadega. Altitudinally, Kombolcha district stretches between 1200 and 2460 meters above sea level and it falls under Woinadega (74%) and Kola (26%) agro-climatic zones. The district has a total population of about 120 000 and more than 90% of the population resides in rural areas. The annual rainfall ranges from 600 mm to 900 mm.

Mixed farming system with the predominance of crop production is practiced in the two districts. Farmers produce different crops in order to secure their family food supply and cover various household
expenditures. Permanent crops include sorghum, maize, haricot bean and groundnut. Vegetables mainly
potato, onion, cabbage, lettuce, tomato, carrot and beet root are commonly produced. In addition, chat (note
2) is the dominant cash crop widely produced in the areas. Keeping animals in their homestead to provide
feed by the cut and carry system is commonly practiced in the highland areas. The production system in the
areas is both rain-fed and irrigated systems the rain-fed production system being most dominant. Unreliable
rainfall, low adoption of modern technologies, lack of credit facilities, shortage of potable water supply,
backward infrastructural facilities, deterioration of forests and soil fertility are the major problems. On the
other hand, varied and favorable climatic conditions are the major potentials of the district.

2.2. Types and Sources of Data
The study was based on both primary and secondary data. Primary data related to the socioeconomic
characteristics of the farmers such as land size, family size, land, labor, crop yield, and livestock owned,
machinery owned, risk sources and mitigation strategies were gathered from the sample farmers. Secondary
data like prices of commodities and other related information were collected from Agriculture and Rural
Development and marketing offices of the districts. Different published and unpublished materials were
also reviewed. In addition, websites were consulted for secondary data and supporting literature.

2.3. Sampling Design and Methods of Data collection
For the study, both probability and non-probability sampling techniques were employed. Haramaya and
Kombolcha districts (note 3) were selected purposively since they have vegetable dominated-mixed
farming systems and smallholder farmers in these districts are exposed to various types of risks. A two
stage random sampling procedure was then employed to select 167 respondents from the districts. In the
first stage, 10 peasant associations (4 from Kombolcha and 6 from Haramaya districts) were selected
proportionately. In the second stage, 167 sample households were selected from each peasant association
using probability proportional to size method. The data for the study were collected with the aid of a
questionnaire. In addition, information was obtained from discussions with District Agricultural Officers
and extension workers and other farmers outside the formal sample to supplement the data.

Before conducting the formal survey, the questionnaire was pre-tested by interviewing seven farmers. This
greatly helped to improve the questionnaire to generate relevant data and to have clear understanding about
the farming system in the areas. The researcher recruited twelve enumerators, who have good knowledge
and experience on farming system and they were given training on the objective of the data collection, how
they approach and clarify questions to farmers and record the data properly.

2.4. Methods of Data Analysis
In order to address the objectives of the study, descriptive analysis and econometric model were employed
in this study.

2.4.1. Descriptive statistics
Descriptive statistics was employed to analyze and present the socioeconomic characteristics of sample
households and to rank farmers’ responses with regard to certain questions. In addition, prioritizing factors
influencing farm household decisions were made using simple descriptive statistical tools. Descriptive
analysis involved calculating of frequency, percentages, means, and standard deviation.

2.4.2. Econometric model
Tobit regression was used to analyze covariates of crop diversification and its intensity. The variable used
to measure crop diversification is a modification of the Entropy index that lies strictly between zero and
one. The diversity index is censored because some of its values cluster at the limit (i.e., 0 for complete
specialization and 1 for perfect diversification). Standard ordinary least squares (OLS) or seemingly
unrelated regression (SUR) of the diversity index will yield biased and inconsistent estimates in this situation. For the same reasons as above, it is not appropriate to use a classical regression model for this purpose. Hence, opting for other econometric models is inevitable for handling the matter. One of such models is censored regression models.

Censored regression models refer to a model in which we observe the dependent variable only if it is above or below some cut off level. Tobit model is a special case of censored regression models that arise when the dependent variable is limited (or censored) from above and/or below. It is a non-linear model which employed maximum likelihood estimation technique which estimates the likelihood of crop diversification and its intensity. The Tobit approach has been applied in previous studies of crop diversification (e.g. Allison 2010). This model is appropriate since the dependent variable is an index which takes values between 0 and 1 inclusive. The dependent variable of the model can be either left-censored, right-censored, or both left-censored and right-censored, where the lower and/or upper limit of the dependent variable can be any number.

The two-limit Tobit model can be specified as:

\[ y_i^* = \beta' x_i + \epsilon_i \]  

where \( y_i^* \) is a latent variable (unobserved for values smaller than 0 and greater than 1) representing specialization or diversification index; \( x_i \) is a vector of explanatory variables; \( \beta' \) is a vector of unknown parameters; and \( \epsilon_i \) is a disturbance term.

Denoting \( y_i \) (crop diversification index) as the observed dependent variable the two limit Tobit model can be specified as:

\[
\begin{align*}
 y_i &= \begin{cases} 
 0 & \text{if } y_i^* \leq 0 \\
 y_i^* & \text{if } 0 < y_i^* < 1 \\
 1 & \text{if } y_i^* > 1 
\end{cases}
\end{align*}
\]  

In principle, a maximum likelihood approach may be employed to address the censoring (e.g., Tobit model) and account for correlations in error terms across equations by specifying a multivariate density function for the error terms (Benin et al. 2004). Censored regression models (including the standard Tobit model) are usually estimated by the Maximum Likelihood (ML) method. The log likelihood function is specified with an assumption that the disturbance term \( \epsilon \) follows a normal distribution with mean 0 and variance \( \sigma^2 \).

Unlike traditional regression coefficients, the Tobit coefficients cannot be interpreted directly as estimates of the magnitude of the marginal effects of changes in the explanatory variables on the expected value of the dependent variable. In a Tobit equation, marginal effect includes both the influence of the explanatory variable on the probability of crop diversification as well as its intensity. More explicitly, the marginal effect takes into consideration that a change in an explanatory variable will affect simultaneously the number of sample farmers diversifying their production and the extent of diversification.

**Definition of Variables and Hypothesis**

**Dependent Variable:** The dependent variable used is crop diversification measured by the modified entropy index. The modified entropy index accounts for both the mix of crops and the relative importance of each crop to the farm business. The index spans a continuous range from 0 to 1; the value of the index for a completely specialized farm producing one crop is 0 whereas a completely diversified farm with equal shares of each crop has an index of 1.

The modified entropy index which represents the dependent variable in this study is obtained using the formula:
Where, $D_{index}$ is crop diversification index, $p_i$ is proportion of area under $i^{th}$ crop. $i$ refers to each of the $N$ possible enterprises.

At maximum diversification, this index takes a value of 1 and at maximum specialization it attains a value of 0. The modified entropy index provides a uniform and fixed scale and hence it is used as a norm to compare and rank the extent of diversification spatially. Other empirical studies on determinants of farm diversification have used the same index (e.g. Mishra & El-Osta 2002).

**Explanatory variables:** The following were expected to be explanatory variables determining the level of farm diversification. These variables are identified based on review of the empirical works and economic theory.

**Farm size:** Farm size is a continuous variable referring to the total land holding by the farmer measured in hectares. It was expected to positively related to crop diversification. Previous findings indicate that diversification activities are concentrated on large farms (Weiss and Briglauer 2000; Mishra and El-Osta 2002; Benin et al. 2004; Fetien et al. 2009). It was, therefore, hypothesized that larger farms may be more diversified.

**Age of the household head:** Age is a continuous variable and a negative relationship was expected between age of the farm operator and diversification. Previous studies suggest that older farm operators are less likely to diversify (Mishra and EL-Osta 2002). One possible explanation is that older farm operators are less risk-averse and less diversified whereas young and beginning farm operators might be more risk-averse.

**Household size:** The size of the farm household was expected to be positively related with diversification. Previous studies also support this hypothesis (Weiss & Briglauer 2000; Mishra & El-Osta 2002; Benin et al. 2004).

**Distance to markets:** Previous works on farm diversification highlighted the importance of proximity to main roads and markets for development of other farm enterprises (Benin et al. 2004). A negative correlation was expected between crop diversification and proximity to markets. It is a continuous variable and measured in kilometers.

**Number of extension contacts:** The larger the number of contacts a farmer has with an extension agent, the more he is likely to engage in production of large number of enterprises. It is a continuous variable and measured as the number of contacts the farmer has with the agent per year. The finding of Fetien et al. (2009) also supported this hypothesis.

**Farm Machinery (tractor, water pump):** Whether the farmer has farm machinery or not was expected to be positive determinant of crop diversification. One study in India has indicated that there is a positive relation between possession of machinery (tractor) and farm diversification (Mishra & El-Osta 2002).

**Off/non-farm income:** This variable is defined for this particular study as all incomes except on-farm income earned by the sample household. Off-farm/non-farm income diversifies a farm operator’s income portfolio and reduces the degree of farm diversification and possibly off-farm work is not compatible with the labor demands of farm diversification. Off-farm income from wages and salaries is a continuous variable that was expected to be negatively correlated with crop diversification. Mishra & El-Osta (2002), Weiss & Briglauer (2000) have also reported inverse relations.

**Number of farm plots:** This variable is a continuous and expected to positively influence crop diversification. The more the number of farm plots, the more is the diversification (Benin et al. 2004).

**Access to market information:** It is a dummy variable taking value 0 if a household have no access to market information and 1 otherwise. This variable was hypothesized to have a positive relation with crop
diversification. **Irrigation intensity**: It is a continuous variable measured as the total irrigated area under each crop divided by the total cropped area by a farmer. The hypothesis is based on Lonnie *et al.* (1989).

**Sex of the household**: This is a dummy variable assuming 0 if female and 1 if male household head. The variable was expected to have a positive effect on diversification due to the fact that there is skill or requirements for frequent and early ploughing. Hence, male headed households are more likely to diversify (Fetien *et al.* 2009).

3. **Results and Discussions**

The major findings of the study are discussed in the following sections. The first section deals with the socio-economic characteristics of the sample farmers in the two districts while the second section describes the patterns and trend of crop diversification in the study area. The final section presents and discusses the covariates of crop diversification using the Tobit regression.

3.1. **Socioeconomic Characteristics of the Sample Households**

The socioeconomic characteristics of the sample households are presented in the Tables 1 and 2. The descriptive statistics results indicate that the majorities (more than 95%) of the households are male-headed and married. About 70% of the sample respondents in Haramaya district are illiterate whereas about 75% of the respondents in Kombolcha district are at least able to read and write. Against this background, it is interesting to note that the number of household heads with no education is clearly higher in Haramaya district than in Kombolcha. This suggests lower investment in human capital among the households of Haramaya district.

About 6% of the sample households in the study area are engaged in off-farm or non-farm income generating activities. This indicates that considerable proportion of the farmers does not have access to off-farm and non-farm income sources. This situation is supposed to constrain the off-farm/non-farm income to serve as a hedge against crop failure risk.

The socio-economic profile of households for the continuous variables is summarized in table 2. The average age of the household heads in the study area is above 40 years and farmers in Haramaya district have slightly longer years of experience in farming than household heads in Kombolcha district. On average, the relatively large household size is maintained in the study areas (around 5 adults per household). Larger household size allows the household flexibility to pool resources and share risks by taking advantage of household return to scale and labor supply they need during peaked demand season.

The average land holding in the study areas is less than the national average which is 1.53 hectares. The average land holding in Haramaya district (0.74 ha) is higher than that of Kombolcha (0.41 ha) on average. In Haramaya district, there is greater variability in land ownership as indicated by the higher standard deviation (0.598). The average livestock holding measured in tropical livestock units is slightly higher in Haramaya (3.05) than in Kombolcha (2.72) district. This might be related to the fact that households in Haramaya district have enough land to raise livestock besides growing crops.

Farmers in Haramaya are expected to walk long distance to reach the nearest market as indicated by the average distance to market. This attributed to the fact that most of the peasant associations of the district are located far from the district capital, Haramaya town. So, transport costs are therefore, potential constraints particularly for the farmers in Haramaya district.

The average annual farm income is relatively higher and more variable in Haramaya district than Kombolcha. The result of the descriptive statistics indicated that the average non/off-farm income was higher in Kombolcha district. However, there is greater variability among households in Kombolcha district than in Haramaya.

3.2. **Crop choices and trend in Crop diversification**
Land allocation is an indicator of importance of the crop under consideration. Crop portfolio composition of the study areas seems to contain significantly higher proportion of low-risk low-return crops. To some extent, on the other hand, they contain low concentration of high value cash crops. Farms in the highlands are commonly producing vegetables and staple crops and production of chat is common in the mid-highlands.

The areas allocated to major crops in the study areas for six production years (2004–2009) are presented in Table 3. The data is obtained from the District Agricultural Offices. The area allocated for all crops remains nearly the same indicating no significant increase in area allocated to both cereal crops and vegetables for the six years. However, the increase in area allocated for cereals seems higher than that of vegetables indicating farmers are tending to be less commercialized through time. Hence, there is a need to revert this tendency.

There appears an increasing trend in diversification in the study areas. As already indicated, analyzing the trend in diversification may help to know risk-aversion behavior of smallholder farmers. As long as crop diversification dominates the literature as a dominant risk management strategy, risk-averse smallholder farmers are characterized by diversified farms. Hence, increase in diversification index with time indicates increase in risk aversion. Crop diversification as measured by land under each crop has shown an increasing trend for the 2004 to 2009 production period. There have been changes in the pattern of agricultural diversification at the district level. Farms keep on diversifying their farms and the usual notion of crop diversification as a risk management practice is also approved in the present study. The study also found that farmers in eastern Hararghe highlands of Ethiopia are tending to be risk-averse over time. The major contributor to increased farm diversification is also believed to be the increase in land fragmentation in the study area. This has important implication to technology adoption behavior of smallholder farmers. With increase in risk aversion through time, farmers might become more reluctant to adopt technologies.

3.3. Pattern of crop diversification

The different forms of crop diversification are crop rotation and intercropping. Intercropping is the most common form of enterprise diversification in eastern Hararghe. The result of the survey indicates that farmers intercrop for three major reasons: to increase soil fertility, better use of resource (land in this case since it is scarce) and for minimizing risk due to loss from another enterprise(s). According to Bezabih and Hadera (2007), the advantage of intercropping may also entail supplementary relationship which calls for physical support of one crop to the other crop and erosion control through providing continuous leaf cover over the ground surface.

Crop rotation is believed to reduce disease incidence and increase soil fertility. Intercropping might allow efficient use of land by growing more crops together. Farm diversification is growing different crops on different plots hence reduce risk of losing crops from all fields. The survey result also indicated that intercropping as a form of diversification plays significant roles. About 78% of the respondents indicate that their reason for intercropping is to reduce risk of losing yield from crop due to specialization. Other reasons for intercropping were shortage of land and to increase soil fertility by intercropping with vegetables and leguminous crops. This result indicates that farmers in the study area are much concerned about land scarcity, soil fertility and risk management.

The proportion of farmers practicing crop rotation and crop rotation, average number of farm plots operated and number of crops grown in a single production period are indicated in Table 4. The result indicated that significant proportion of the sample farmers practice crop rotation and intercropping. The major crop rotation practices are sorghum between two vegetable crops and vegetables between maize years. The result related to the average number of farm plots operated by a household indicates that a household on average operates about 4 farm plots. This indicates that farm land fragmentation is persistent in the area. A typical farmer grows 4 crops in a single production period. If number of enterprises grown by a farmer is taken as a measure of crop diversification, it tells that the farmers are diversifying their production.
3.4. Determinants of Farm Diversification

The results of a Tobit model which analyzed covariates of crop diversification were estimated using STATA Version 9.10 statistical package. The results are summarized and presented in Table 5.

Multicollinearity was checked using variance inflation factor (VIF) for continuous variables and Contingency Coefficients for dummy variables. The calculated VIF values are all less than 10 (the cut-off point) and contingency coefficients were less than 0.75 (the cut-off point) which indicated that multicollinearity is not a serious problem. Since the Tobit model has a probit component and its results are sensitive to the assumption of homoscedasticity, a robust standard error Tobit regression was run.

The coefficient of livestock ownership as is negative and significant at 5% indicating an inverse relationship between livestock ownership and crop diversification. The explanation for the result is, livestock as measure of wealth may act as insurance against crop production risk, bearing a negative relationship with crop diversification. So, households with large number of livestock are less likely to grow more crops. The result is consistent with the findings of Benin et al. (2004), but in contrast to that of Feiten et al. (2009).

The positive coefficient for the number of farm plots operated by a household indicates that households with more number of farm plots are more likely to diversify by growing different crops on each plot of land. The variable is significant at 1% probability level. More fragmented farms with larger numbers of farm plots have more crops that are likely to be grown more evenly since the farm plots are approximately of equal size. Likewise, by diversifying plot locations across the soil toposequence, farmers are able to reduce plot-yield co-variation and thus reduce aggregate production variability. The result of the study is in line with the finding of Berhanu & Moti (2010) who found land fragmentation to be most important determinant of crop diversification. An increase in farm plots will increase the intensity of diversification by 1.40% on average.

Access to market information significantly affects crop diversification at 10% probability level. Households having access to market information are more likely to diversify their production since they have the information related to supply, demand and prices of most crops.

It is evident from the result that households who own farm machinery (water pump) are more likely to diversify because they can properly perform different farming operations on time and can market their produce easily. The variable is significant at 5% probability level and the result is consistent with the finding of Muhammed et al. (2008). Owning machinery (water pump) is related to an increase in probability of crop diversification by 0.29% and the intensity by 3.85% on average.

There appears a negative and significant relationship between number of extension contacts per year and crop diversification and the coefficient is significant at 5% probability level. This might be associated with the extension system which is concentrated on enhancing farmers’ productivity and profitability which favors specialization at micro-level and overlooks the role of crop diversification in risk minimization. The result is consistent with the finding of Fetien et al. (2009) and in line with the explanation of Richard (1998) who stated that extension contact has discouraged intercropping for a number of years and has promoted the growing of pure crops targeted for commercial purposes. The result is also supported by Ayinde et al. (2008) where positive relationship was found between risk coefficients of the household and their access to extension services.

Irrigation intensity measured as the proportion of area irrigated was found to significantly and positively affect crop diversification at 10% probability level. Farmers who have access to irrigation have opportunities to grow more crops. This is observed in the study area where farmers having access to irrigation grow vegetables on their farms. The result of this study is in line with the explanation of Lonnie et al. (1989) who found positive relationship between irrigation and enterprise diversification.

Finally, geographical location of farms determines cropping pattern, rainfall, soil productivity and so on. To this end, in addition to the diversity with respect to Peasant associations within the sample districts, a district dummy was defined and included in the model. The result indicates that compared to farmers in Kombolcha district (northern part), farmers in Haramaya district (North central part) are more likely to
diversify their farms. However, this difference may be attributed to variables which are not important for classifying the two districts as having different farming systems or the differentiating factor might be simply location of farms in the study area.

4. Conclusion and Policy Issues

4.1. Conclusion

Considering the multidimensional importance of crop diversification as a risk management strategy and an engine for ensuring food security, the present study was conceived to assess the patterns, trend and determinants of crop diversification at farm level.

Large number of farmers practices crop rotation and intercropping for different important reasons which include reducing disease and pest infestations, manage with land scarcity, improving soil fertility and reducing risks. There appears an increasing trend in crop diversification which attributes to increase in area allocated to cereal crops. This indicated that farmers are tending to be more risk-averse and this will have an implication in technology adoption and agricultural development in the area.

A modified entropy index was considered as an approach to measure crop diversification based on area in hectare under each crop. The Tobit estimates indicates that access to market information, irrigation intensity, machinery ownership, livestock size, extension contact, number of farm plots and location of farms are found to be the most important factors that significantly influence crop diversification.

4.2. Policy Implications and Further Research

Based on the empirical evidences, the following recommendations are suggested to promote crop diversification which is a mechanism for farmers for avoiding risks and to ensure a sustainable level of farm returns.

- Policy makers need to devise a way to revert the farmers’ tendency to become more risk-averse through time. This might improve farmers’ technology adoption decisions.
- Attention should be given to design strategies so that farmers might diversify their production to high value crops leaving small land for staple cereal crops.
- Farm machinery especially water pumps should be provided through easy loan schemes and investment in irrigation projects would also save the farmers from drought and erratic rainfall.
- Access to market information need to be given attention. Infrastructures like farm to market roads and access to markets can play positive role in enhancing diversification among farmers.
- The extension system should not stick only to increasing productivity or profitability; risk reduction should be part and parcel of an extension system. Negative relation between diversification and extension contact urges that production under risk should be made part of farmers’ training and extension programs brought in to improve the efficiency of individual farms.
- Diversification of farming is very useful investment to mitigate risk. Inclusion of livestock enterprises might minimize the risks in the farming. In addition, it might contribute to efficient use of labor. Hence, future research in the areas should integrate livestock in the model.
- Even though crop diversification is thought to be dominant risk management strategy, further study regarding farmers’ perception about it, its costs and benefits need to be explicitly studied.

5. References


**Notes**

Note 1. Kola (lowland) agro-climatic zone (500-1500 m.a.s.l.) is characterized by annual mean rainfall of 410-820 mm and temperature of 20-25°C. Woinadega (mid-highland) agro-ecological zone (1500-2300 m.a.s.l.) is characterized by an average annual rain fall of 600-2000 mm and mean temperature of 15-20°C (Hurni 1995 cited in Belaineh 2003).

Note 2. Chat or khat (Catha edulis) is cultivated as cash crop for its young leaves and tender stems chewed as a mild natural stimulant, which like coffee or alcoholic beverages, plays an important role in the social life of people in Ethiopia in general and in Hararghe and Somali region in particular ( Klingele 1998).
Note 3. Ethiopia is a federal state of regions. Every region is administratively structured into zones and zones are again divided into weredas which are called districts. Every wereda is divided into Peasants Association, which is the lowest administrative unit. The peasant associations are called Kebeles in urban areas.

Table 1: Summary statistics of the sample households (categorical variables) (%)

<table>
<thead>
<tr>
<th>Variables</th>
<th>Haramaya</th>
<th>Kombolcha</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex of household head</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Male</td>
<td>97.9</td>
<td>92.5</td>
<td>95.4</td>
</tr>
<tr>
<td>Female</td>
<td>2.1</td>
<td>7.5</td>
<td>4.6</td>
</tr>
<tr>
<td>Marital Status of household head</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Single</td>
<td>3.2</td>
<td>1.3</td>
<td>2.3</td>
</tr>
<tr>
<td>Married</td>
<td>96.8</td>
<td>98.7</td>
<td>97.7</td>
</tr>
<tr>
<td>Educational level of household head</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Illiterate</td>
<td>69.5</td>
<td>25</td>
<td>49.1</td>
</tr>
<tr>
<td>Elementary/Primary</td>
<td>28.4</td>
<td>71.25</td>
<td>48.1</td>
</tr>
<tr>
<td>High School and above</td>
<td>2.1</td>
<td>3.75</td>
<td>2.8</td>
</tr>
</tbody>
</table>

Table 2: Summary statistics of the sample households (continuous variables)

<table>
<thead>
<tr>
<th>Variables</th>
<th>Haramaya</th>
<th>Kombolcha</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age of head (Years)</td>
<td>42.18</td>
<td>37.54</td>
<td>40.06</td>
</tr>
<tr>
<td>Farming Experience (Years)</td>
<td>21.51</td>
<td>16.96</td>
<td>19.39</td>
</tr>
<tr>
<td>Family Size (Persons)</td>
<td>6.03</td>
<td>6.74</td>
<td>6.35</td>
</tr>
<tr>
<td>Family Size (Adult Equivalent)</td>
<td>4.87</td>
<td>5.08</td>
<td>4.97</td>
</tr>
<tr>
<td>Farm Size (hectares)</td>
<td>0.74</td>
<td>0.41</td>
<td>0.59</td>
</tr>
<tr>
<td>Distance to market (Km)</td>
<td>7.9</td>
<td>3.28</td>
<td>5.73</td>
</tr>
<tr>
<td>Distance to farm (Km)</td>
<td>1.051</td>
<td>0.841</td>
<td>0.899</td>
</tr>
<tr>
<td>Livestock Owned (TLU)</td>
<td>3.05</td>
<td>2.79</td>
<td>2.89</td>
</tr>
<tr>
<td>Farm Income (Birr)</td>
<td>12050.46</td>
<td>8,541.4</td>
<td>10201.20</td>
</tr>
<tr>
<td>Non/Off farm Income (Birr)</td>
<td>144.04</td>
<td>247.95</td>
<td>193.76</td>
</tr>
</tbody>
</table>

Table 3: Proportion of area allocation and diversification index for the 2004 – 2009 production period

<table>
<thead>
<tr>
<th>Crops</th>
<th>2004</th>
<th>2005</th>
<th>2006</th>
<th>2007</th>
<th>2008</th>
<th>2009</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maize</td>
<td>0.183</td>
<td>0.183</td>
<td>0.182</td>
<td>0.189</td>
<td>0.182</td>
<td>0.188</td>
</tr>
<tr>
<td>Sorghum</td>
<td>0.139</td>
<td>0.138</td>
<td>0.139</td>
<td>0.166</td>
<td>0.146</td>
<td>0.154</td>
</tr>
<tr>
<td>Potato</td>
<td>0.049</td>
<td>0.063</td>
<td>0.0634</td>
<td>0.055</td>
<td>0.060</td>
<td>0.058</td>
</tr>
<tr>
<td>Onion</td>
<td>0.051</td>
<td>0.026</td>
<td>0.033</td>
<td>0.048</td>
<td>0.059</td>
<td>0.041</td>
</tr>
<tr>
<td>Beet root</td>
<td>0.0057</td>
<td>0.0074</td>
<td>0.0079</td>
<td>0.012</td>
<td>0.015</td>
<td>0.0071</td>
</tr>
<tr>
<td>Cabbage</td>
<td>0.0056</td>
<td>0.0052</td>
<td>0.0057</td>
<td>0.012</td>
<td>0.0137</td>
<td>0.0094</td>
</tr>
<tr>
<td>Carrot</td>
<td>0.0050</td>
<td>0.0044</td>
<td>0.0057</td>
<td>0.013</td>
<td>0.016</td>
<td>0.0051</td>
</tr>
<tr>
<td><strong>Diversification Index</strong></td>
<td>0.439</td>
<td>0.427</td>
<td>0.437</td>
<td>0.496</td>
<td>0.493</td>
<td>0.462</td>
</tr>
</tbody>
</table>
Table 4: Pattern of diversification in the study area

<table>
<thead>
<tr>
<th>Particulars</th>
<th>Mean</th>
<th>Stand. Dev.</th>
<th>Maximum</th>
<th>Minimum</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of farm plots</td>
<td>3.26</td>
<td>1.68</td>
<td>11</td>
<td>1</td>
</tr>
<tr>
<td>Number of enterprises</td>
<td>4.30</td>
<td>1.43</td>
<td>8</td>
<td>2</td>
</tr>
</tbody>
</table>

Proportion of Farmers (%)

<table>
<thead>
<tr>
<th></th>
<th>Yes</th>
<th>No</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercropping</td>
<td>78.9</td>
<td>21.1</td>
</tr>
<tr>
<td>Crop rotation</td>
<td>54.4</td>
<td>45.6</td>
</tr>
</tbody>
</table>

Table 5: Tobit regression results, determinants of farm diversification

<table>
<thead>
<tr>
<th>Variables</th>
<th>Coefficients</th>
<th>Std. Error</th>
<th>ME(Probability)</th>
<th>ME(Intensity)</th>
</tr>
</thead>
<tbody>
<tr>
<td>District</td>
<td>0.0557*</td>
<td>0.0323</td>
<td>0.0021</td>
<td>0.03634</td>
</tr>
<tr>
<td>Access to Market Information</td>
<td>0.0702*</td>
<td>0.0413</td>
<td>0.0038</td>
<td>0.04913</td>
</tr>
<tr>
<td>Farm size (hectares)</td>
<td>-0.0137</td>
<td>0.0122</td>
<td>-0.00004</td>
<td>-0.00889</td>
</tr>
<tr>
<td>Irrigation Intensity</td>
<td>0.1694**</td>
<td>0.0779</td>
<td>0.0036</td>
<td>0.10965</td>
</tr>
<tr>
<td>Machinery Ownership</td>
<td>0.0588**</td>
<td>0.0279</td>
<td>0.0029</td>
<td>0.03849</td>
</tr>
<tr>
<td>Sex of household head</td>
<td>0.0515</td>
<td>0.0571</td>
<td>0.0013</td>
<td>0.03557</td>
</tr>
<tr>
<td>Market distance to District (Km)</td>
<td>-0.0027</td>
<td>0.0035</td>
<td>-0.00015</td>
<td>-0.00172</td>
</tr>
<tr>
<td>Age of household head (Years)</td>
<td>-0.0006</td>
<td>0.0013</td>
<td>-0.000029</td>
<td>-0.00040</td>
</tr>
<tr>
<td>Number of Extension Contacts</td>
<td>-0.0019**</td>
<td>0.0008</td>
<td>-0.000038</td>
<td>-0.00123</td>
</tr>
<tr>
<td>Distance from farm to home (Km)</td>
<td>-0.0035</td>
<td>0.0133</td>
<td>-0.000059</td>
<td>-0.00224</td>
</tr>
<tr>
<td>Number of Farm plots</td>
<td>0.0217***</td>
<td>0.0072</td>
<td>0.00072</td>
<td>0.01408</td>
</tr>
<tr>
<td>Livestock holding (TLU)</td>
<td>-0.0123**</td>
<td>0.0047</td>
<td>-0.00042</td>
<td>-0.0000</td>
</tr>
<tr>
<td>Education level of head (Year)</td>
<td>-0.0047</td>
<td>0.0043</td>
<td>-0.00024</td>
<td>-0.00307</td>
</tr>
<tr>
<td>Household size (Adult Equivalent)</td>
<td>-0.0086</td>
<td>0.0078</td>
<td>-0.00034</td>
<td>-0.00557</td>
</tr>
<tr>
<td>Farming Experience (Years)</td>
<td>-0.0017</td>
<td>0.0016</td>
<td>-0.000039</td>
<td>-0.00109</td>
</tr>
<tr>
<td>Constant</td>
<td>0.7830***</td>
<td>0.0913</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

LR $\chi^2$ (16) = 36.71
Log likelihood = 72.20***
Number of observations = 167

* Statistically significant at 10% level; ** statistically significant at 5% level; *** statistically significant at 1% level respectively. ME = marginal effects
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