

Socio-Economic and Institutional Factors Influencing Adoption of Agroforestry in Arid and Semi Arid (ASALs) Areas of Sub Saharan Africa

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Abstract: The slow pace in adoption of agroforestry in the Sub Saharan Africa, particularly in arid and semi arid areas (ASALs) warrant an understanding of the underlying factors. Several studies have indicated that in resource limited countries including the ASALs, socio-economic and institutional factors may play a key role in adoption of agricultural production and technologies. Therefore, this study evaluated 16 socioeconomic and institutional factors influencing the adoption of agroforestry in Eastern Kenya within the Sub Saharan Africa. The study used survey research design. Based on a sample of 248 household heads, we gathered data using questionnaires and interviews. The influence of socio-economic factors on adoption was performed using binary logistic regression model. Our results showed that agroforestry was adopted by 82% of the respondents. The main agroforestry practices adopted by farmers were boundary planting (73.8%), hedgerow (69.4%), woodlots (53.2%), scattered planting (51.2%) and alley cropping (37.1%). The levels of education, household size, access to credit, and training were significant ($P < 0.05$) factors influencing the adoption of agroforestry. More resources in the forms of access to credit, inputs and training on agroforestry adoption should be high priority focus to enhance adoption of the practice.

Keywords: ASALs, Agroforestry adoption; Socio-economic factors; Institutional factors; Machakos, Kenya

1. INTRODUCTION

Due to high productivity and sustainable land use, adoption of agroforestry is ubiquitous at the global scale (Dalemans *et al.* 2018; Fleming *et al.* 2019; McAdam and Curran 2018). Studies have shown that agroforestry practices slow or reverse land degradation, sequester carbon from the atmosphere and secure rural livelihoods through provision of economic benefits such as increase food security in rural areas (Catacutan *et al.* 2017; Montagnini and Metzler 2017; Saqib *et al.* 2019; Sharma and Sharma 2017; Waldron *et al.* 2017). Trees managed by farmers also provide ecosystem services and functions in addition to the products and services that motivate farmers to plant them (Fagerholm *et al.* 2016; Kuyah *et al.* 2017). These services are of particular importance in many low-income countries in Africa, where large proportions of the populations work in an agricultural sector that does not attract much investment from either government or private sector (Benjamin and Sauer 2018; Meijer *et al.* 2015). The contribution of agroforestry to livelihood especially in the rural areas has witnessed an increased recognition by practitioners (Munsell *et al.* 2018), and international bodies such as the United Nations and World Bank, ICRAF, government and non-governmental organizations (NGOs) for global adoption of agroforestry (Ajayi and Place 2012; Place *et al.* 2012). This advocacy has resulted in approximately 350 million agroforestry adopters, who dedicate at least 5 to 10% of their farms to practice agroforestry (Pattanayak *et al.* 2003).

Although there has been significant advances in research on agroforestry adoption over the past three to five decades, it is agreeable that adoption of agroforestry including agroforestry practices and technologies have lagged behind the scientific and technological advances in agroforestry research (Dalemans *et al.* 2019; Kabwe *et al.* 2016; Mwase *et al.* 2015). This is particularly true, in the developing countries where agroforestry has lagged behind in its contribution to agricultural productivity, ecosystem services, and human well-being (Miller *et al.* 2017) as compared to the

developed countries in Europe and North America (Brockington *et al.* 2016; Brown *et al.* 2018; Kalaba *et al.* 2010; Sangeetha *et al.* 2016). The underlying factors behind these differences are currently being exploited with broad spectrum of suggestions. Interest in the adoption and practice of agroforestry has increased among smallholder farmers in the developing countries especially in the Sub Saharan Africa (Garrity 2004; Owombo *et al.* 2018). An active area of research therefore concerns the preconditions that must be met for successful establishment of agroforestry. A major research frontier in agroforestry science is the extrapolation of the influences of locally successful practices, to aid in better understanding of barriers to adoption.

There are numerous studies that have looked at the importance of social and economic factors among households (Alavalapati *et al.* 2001; Franzel *et al.* 2002; Matata *et al.* 2010; Zerihun *et al.* 2014). There is also increasing recognition that institutions that support agroforestry as well as the institutional factors may have an impact on the adoption of agroforestry among the rural populations (Binam *et al.* 2017; Mercer 2004). However, there is less emphasis on how combination of socio-economic and institutional factors affect adoption of agroforestry (Alavalapati *et al.* 2001; Franzel *et al.* 2001; Matata *et al.* 2010; Mercer 2004; Mwase *et al.* 2015). This is particularly lacking in the Sub Saharan Africa where there are numerous constraints to adoption of agroforestry. Therefore the contribution of both socio-economic factors and institutional factors on adoption of agroforestry need to be understood in the local context to better understand the barriers to adoption of agroforestry. Therefore the aim of this study was to model 16 socio-economic and institutional factors affecting adoption of agroforestry in semi arid region of Machakos County in Kenya within the Sub Saharan Africa.

2. MATERIALS AND METHODS

2.1. Study Area

The study was conducted in Machakos County (Fig. 1) in Kenya. The county covers an area of 5,953 km². It lies between latitudes 0°45'South and 1°31'South and longitudes 36°45'East and 37°45'East. Most of the land is semi-arid with population of 1,098,584 as per the 2009 Kenya National census (Kenya National Bureau of Statistics 2010). Administratively the county is divided into 11 divisions: Kalama, Kangundo, Kathiani, Machakos Central, Masinga, Matungulu, Mavoko, Mwala, Ndithini, Yathui and Yatta. In terms of political structure, the county has eight constituencies including: Kangundo, Kathiani, Machakos Town, Masinga, Matungulu, Mavoko, Mwala and Yatta. Division and constituency is sometimes referred to as sub-counties. Agroforestry is practiced in Kathiani, Mavoko and Machakos Town. Four sites selected in this study were: Mua Hills (Mavoko, Machakos Town and Kathiani) and Iveti Hills (Machakos Central and Kathiani), Kima-Kimwe and Kalama in Machakos Town Constituency.

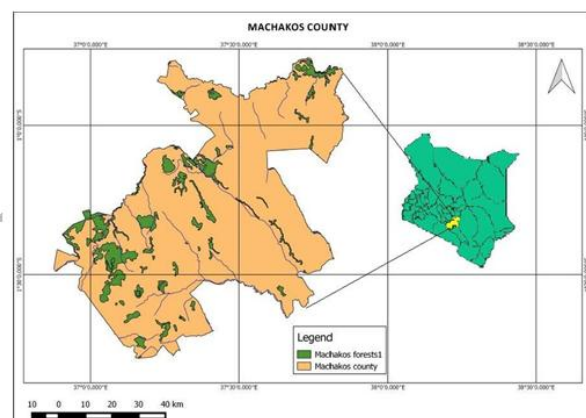


Fig1. Map of Machakos County showing the study area

The local climate is semi-arid with hilly terrain and an altitude of 1000 to 2100 m above sea level. The area is composed of hilltops rising to 1594-2100 m above sea level. The annual average rainfall is 1000 mm (range 500 to 1300 mm) with a bimodal pattern. Temperatures range between 18.7°C and 29.7°C. The soils are shallow dark red volcanic on hilltops and clay soils in the plains. Crop such as maize, beans, pigeon peas, vegetables are dominant. Dairy and beef cattle, sheep and goats are the major livestock kept.

2.2. Research Design

This study was conducted using survey design. Surveys are normally used to systematically gather factual quantifiable information necessary for decision-making (Nardi 2018). Surveys are efficient methods of collecting descriptive data regarding the characteristics of populations, current practices and conditions or needs. Survey study research design was adopted in this study to capture descriptive data from selected samples and generalize the findings to the populations from which the sample was drawn.

2.3. Target Population

The study targeted household heads from Mua Hills (Mavoko, Machakos Town and Kathiani), Iveti Hills (Machakos Central and Kathiani), Kima-Kimwe and Kalama Hills in Machakos Constituency. The number of farmers practicing agroforestry in the region has not been established.

2.4. Sample Size and Sampling Design

Since the actual population was not easy to determine due to changes in the rate of adoption with respect to time, the sample size utilizing proportion of the households adopting agroforestry (Nzilu 2015) was used. According to Nzilu, 80% of the households had adopted agroforestry in Mwala Constituency (Machakos County). The appropriate sample size was therefore computed using the formula described

in (Mugenda and Mugenda 2003) as:
$$n = \frac{z^2 p(1-p)}{d^2}$$

Where: n = the desired sample size

z = the z score at the required confidence level $\alpha = 0.05$ (1.96)

p = the proportion in the target population assumed to be adopters (0.8)

d = permissible marginal error (the level of statistical significance, set at $\alpha = 0.05$).

Using the values of z , p and d , the value of n was computed as follows

$$n = \frac{1.96^2 \times 0.8(1-0.8)}{0.05^2} = 246$$

The sample size was 246 in addition to information obtained from 2 research assistants who hailed from the region giving a total of 248 respondents. Samples were selected through stratified, random sampling at each of the selected spatial units and used to identify the adopters and non adopters. Adopters were households practicing any form of agroforestry.

2.5. Research Instruments and Data Collection

This study relied on primary data. Data on socio-economic and institutional factors affecting adoption of agroforestry were collected using structured researcher administered questionnaires. The designing of the instruments were such that they endeavored to ensure an in-depth exploration of personal views, feelings and opinions on agroforestry and benefits accrued.

Field surveys of agroforestry adoption were conducted for three months among the selected group of respondents. Identification of agroforestry adopters was conducted by field observation of the households practicing any form of agroforestry. Before data collection, the respondents were contacted in advance and asked to organize their time for the research. Two research assistants were recruited and trained to aid in the collection of data. The questionnaires were administered by physical drop and pick by the researcher and two research assistants. The researcher personally administered the instrument. The researcher made prior visits to assist in defining timings and distribution of the research instruments.

2.6. Validity and Reliability of the Research Instruments

The researcher developed the research instruments based on study aim and the related literature. The salience of the instruments was sought through expert judgment. This was to purposely ascertain the item's construct and content validity and to establish whether the numbers of items are adequate for the purpose intended research.

The reliability of instruments was established through a pilot study in 12 households who did not participate in this study. The results of the study were used to compute the reliability of the instruments. Cronbach’s coefficient alpha was used to determine the reliability of the instruments. The study considered the instrument reliable and acceptable if the computation yielded a reliability coefficient of 0.7 and above. For this study, the reliability coefficient was 0.85 which was determined to be suitable for the research.

2.7. Statistical Analyses

All questionnaire data were coded into Statistical Package for Social Sciences (SPSS 23.0) for analysis. To test the socio-economic and institutional factors influencing adoption of agroforestry binary logistic regression model was used in testing the probability of dichotomous outcome (in our study, adoption or non-adoption) related to a set of independent explanatory variables that are hypothesized to affect the dependent variable (Harrell 2015). The logistic regression model characterizing agroforestry adoption was specified using the formula: $\ln[P_i/(1-P_i)] = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots + \beta_k X_{ki}$

where subscript i is the observation in the ⁱth sample, P is denotes probability of the outcome, β_0 is the intercept, $\beta_1, \beta_2, \dots, \beta_k$ are coefficients associated with each explanatory variable X_1, X_2, \dots, X_k . The coefficients $\beta_1, \beta_2, \dots, \beta_k$ reflect the effect of individual explanatory variables on its log of odds $\ln[P/(1-P)]$. A positive coefficient means that the log of odds increases in tandem with an increase in the corresponding independent variable (Cox 2018). Accordingly, if the log of odds $\ln[P/(1-P)]$ is positively (or negatively) related to an independent variable, both odds $\ln[P/(1-P)]$ and P of the outcome are also positively (or negatively) related to that variable. Nevertheless, this relationship is linear for the log of odds and nonlinear for odds and probability of the outcome. The significance of the variables in the binary logistic regression was tested using Wald statistics. All analyses were declared significant at $P < 0.05$.

The independent (X_i) variables involved in the logistic regression model for agroforestry adoption are defined in Table 1. The summary statistics of the independent variables (X_i) in the logistic regression are presented in Table 2.

Table1. Description of explanatory variables used in the agroforestry binary logistic model of adoption model

Variable	Description
Age (X_1)	Age in years
Gender (X_2)	Value 1 if the respondent is male, 0 otherwise
Marital status (X_3)	Value 1 if the respondent is married, 0 otherwise
Level of education (X_4)	Index for levels of education: Value 1 = None; 2 = Primary; 3 = Secondary; 4 = Tertiary
Household size (X_5)	Number of people in the household
Land size (X_6)	Land size in acres
Location (X_7)	Household residential areas: Index for location 1= Mua Hills; 2 = Iveti Hills; 3 = Kiima Kimwe Hills; 4 = Kalama
Occupation of the household head (X_8)	Value 1 if the respondent is a farmer, 0 otherwise
Farm household income (X_9)	Amount of income earned by the respondents from the farms (US \$)
Non farm household income (X_{10})	Amount of income earned by the respondents not from the farms (US \$)
Access to extension services	Value 1 if the respondent had access to information, 0 otherwise
Access to credits (X_{12})	Value 1 if the respondent had access to credits, 0 otherwise
Access to formal AF training	Value 1 if the respondent had access to agroforestry training, 0 otherwise
Access to information from conservancy groups (X_{14})	Value 1 if the respondent had access to information from conservancy groups, 0 otherwise
Access to inputs from conservancy groups (X_{15})	Value 1 if the respondent had inputs from conservancy groups, 0 otherwise
Frequency of extension visits (X_{16})	Index for extension visits: Value 1 = None; 2 = Rarely; 3 = Yearly; 4 = Monthly; 5 = Often

Table2. Characteristics of agroforestry ‘adopter and non-adopter used in the logistic regression model

Variables	Agroforestry adopters (n = 204)	Agroforestry non adopters (n = 44)
Age (years)	51.2 ± 12.4	49.2 ± 11.4
Gender	0.42 ± 0.12	0.94 ± 0.23
Marital status	0.95 ± 0.22	0.88 ± 0.33
Level of education	8.74 ± 3.01	8.57 ± 3.92
Household size	6.97 ± 2.64	6.15 ± 2.49
Land size	2.70 ± 1.93	2.35 ± 1.67
Occupation of the household head	0.91 ± 0.28	0.88 ± 0.33
Farm household income (US \$ pm)	290.62 ± 22.83	228.25 ± 16.82
Non-farm household income (US \$ pm)	350.02 ± 36.02	96.37 ± 18.47
Access to extension services	0.43 ± 0.12	0.16 ± 0.02
Access to credits services	0.67 ± 0.24	0.03 ± 0.02
Access to formal AF training	0.35 ± 0.07	0.12 ± 0.02
Access to information from conservancy groups	0.62 ± 0.12	0.03 ± 0.01
Access to inputs from conservancy groups	0.15 ± 0.04	0.04 ± 0.02
Frequency of extension visits	1.69 ± 1.14	1.25 ± 0.21

3. RESULTS

The socio-economic profile of the respondents in Machakos County of Kenya during the study is shown in Table 3. Both adopters and non-adopters were dominated by those aged 36-55 years as well as those aged over 55 years. Our sample had proportionally more females than males regardless of the status of agroforestry adoption. Most of the respondents were married. Educational level for majority of the respondent was primary and secondary levels. In terms of household size, majority of the adopters of agroforestry had household size ranging between 6-10 family members which were higher than the non-adopters. The land size ranged between 0.4 to 24 acres where the majority of the households had land size ranging between 2-5 acres followed by those with less than 2 acres. The annual household farm income for 75% of the respondents ranged between US \$ 50 to 5000 while non farm income for majority of the respondents was often below US \$ 50 followed by income levels between US \$ 50 to 100.

Table3. General characteristics of agroforestry ‘adopter and non-adopter

Variable name	Response category	Agroforestry adopters		Agroforestry non adopters	
		Frequency (n = 204)	Percent	Frequency (n = 44)	Percent
Age (years)	18-25	11	5.4	6	6.9
	26-35	28	13.7	8	18.2
	36-55	84	41.2	14	31.8
	> 55	81	39.1	16	36.4
Gender	Female	116	56.9	26	59.1
	Male	88	43.1	18	40.9
Marital status	Single	12	5.9	1	2.3
	Married	192	94.1	43	97.7
Level of education	None	5	2.5	7	15.9
	Primary	112	54.9	18	40.9
	Secondary	73	35.8	14	31.8
	Tertiary	14	6.8	5	11.4
Household size	< 3	3	1.5	0	0.0
	3-5	75	36.8	27	61.4
	6-10	105	51.5	17	38.6
	>10	21	10.3	0	0.0
Land size	< 2 acre	72	35.3	14	31.8
	2-5 acres	106	52.0	26	59.1

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	5.1-10 acres	26	12.7	4	9.1
Farm income (US \$ pm)	< 50	18	8.8	3	6.9
	50-100	53	26.0	14	31.8
	101-200	43	21.1	14	31.8
	201-500	57	27.9	11	25.0
	>500	33	16.2	2	4.5
Non-farm income (US \$ pm)	< 50	68	33.3	17	38.6
	50-100	46	22.5	10	22.7
	101-200	29	14.2	7	15.9
	201-500	30	14.7	10	22.8
	>500	31	15.2	0	0.0

The study established that 82.3% of the respondents adopted agroforestry practices while 17.7% were non adopters. The types of agroforestry practiced by the adopters are shown in Fig. 2. Majority of the respondents adopted boundary planting (73.8%), hedgerow (69.4%), woodlot (53.2%) and scattered planting (51.2%) while alley cropping was the least preferred agroforestry practice (37.1%).

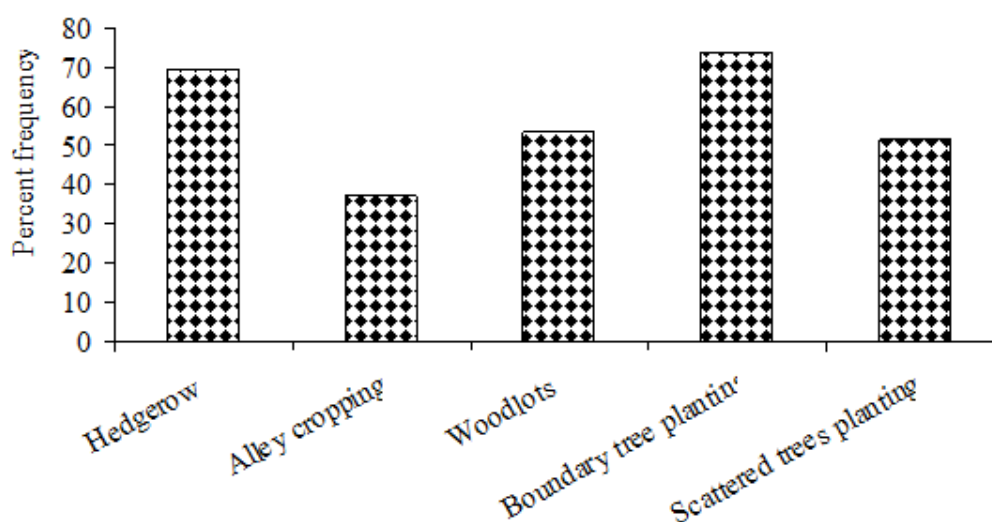


Fig2. Types of agroforestry practiced by the local community members who adopted the practice

The result of the binary logistic regression showing the relationship between 16 socioeconomic and institutional factors on adoption of agroforestry practices are shown in Table 4. The variables in the equation was described using $B = 1.534$, $SE = 0.166$, $Wald = 85.161$, $P = 0.0000$ and $Exp(4.636)$. The significant factors explaining the determinants of agroforestry adoption were levels of education, household size, access to credit and access to training.

Table4. Binary logistic regression showing the relationship between 16 socioeconomic and institutional factors on adoption of agroforestry practices

Variables in the equation	Coefficient	S.E.	Wald	df	P-value
Age	0.248	0.231	1.151	1	0.283
Gender	-0.081	0.404	0.04	1	0.841
Marital status	-1.608	1.143	1.98	1	0.159
Level of education	1.379	0.301	5.588	1	0.021*
Household size	1.219	0.392	9.679	1	0.002**
Land size	-0.561	0.333	2.831	1	0.092
Location	-0.511	0.321	2.674	1	0.095
Occupation of the household	-0.001	0.642	0.0043	1	0.998
Farm income	0.261	0.175	2.221	1	0.136
Non farm income	0.059	0.151	0.151	1	0.697
Access to extension	-1.001	0.616	2.641	1	0.104

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Access to credit	2.616	0.8	10.686	1	0.001***
Access to training	1.682	0.844	3.974	1	0.046*
Access to information from conservancy group	-0.211	0.121	1.611	1	0.595
Access to inputs from conservancy group	0.204	0.221	1.131	1	0.183
Frequency of extension visits	0.073	0.33	0.048	1	0.826
Constant	-1.752	1.786	0.962	1	0.327

* $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$

Adoption of agroforestry among household members and significant socio-economic and institutional factors are shown in Table 5. The study established that level of education was the significant factor and individual were 5.588 times likely adopt agroforestry than those without any education. In this study majority of the adopters were in the range of primary and secondary levels of education. Household size was 9.679 more likely to affect adoption of agroforestry where, most adopters of agroforestry were from large household size with 6-10 people. Access to credit was 10.686 times more likely to affect adoption where it was established that at least 8.8% of the adopters had access to credit while none of the non adopters accessed any form of credit. Finally household with access to formal training were 3.974 more likely to adopt agroforestry than those without access to training. We established that at least 28% of the adopters had access to formal training on agroforestry compared to only 4.5% of the non adopters.

Table5. Relationships between the significant socio-economic status and institutional factors and adoption of agroforestry in Machakos County

		% agroforestry adopter (n = 204)	% agroforestry non adopters (n = 44)
Level of education	None	2.5	15.9
	Primary	54.9	40.9
	Secondary	35.8	31.8
	Tertiary	6.9	11.4
Household size	< 3	1.5	0.0
	3-5	36.8	61.4
	6-10	51.5	38.6
	>10	10.3	0.0
Access to credit facilities	Yes	8.8	0
	No	91.2	100
Access to formal AF training	Yes	27.9	4.5
	No	72.1	95.5

4. DISCUSSION

During the study, a total of 82.3% of the respondents adopted agroforestry practices which concurs with previous studies on adoption of agroforestry in similar regions (Makori 2017; Maluki *et al.* 2016; Nzilu 2015; Rotich *et al.* 2017), and appear to be higher than in other countries of Sub Saharan Africa (Ashiagbor *et al.* 2018; Franzel *et al.* 2001; Oloyede and Ayinde 2016). It is possible that the region being in humid dryland, encourages farmers to adopt the practices as ecosystem services derived from natural forests are not available. Most farmers adopted boundary planting (73.8%), hedgerow (69.4%) and scattered planting (51.2%) while alley cropping was the least preferred agroforestry practice (37.1%). This concurs with other studies that have indicated that farmers prefer hedgerow agroforestry which provides shelter, prevents frosts and act wind breaks (Kuyah *et al.* 2016; Lasco *et al.* 2014; Mbow *et al.* 2014; Mitigation 2010). A number of the farmers also adopted boundary planting as wind breakers and to demarcate boundaries of the farmers perhaps in order to avoid trespassers.

Binary logistic regression was chosen because to test the influence of the 16 socioeconomic and institutional factors on the adoption of agroforestry due to its ability to utilize both the continuous and categorical variables and or if they are not nicely distributed (Frölich 2006). When the 16 socioeconomic and institutional factors on adoption of agroforestry practices were tested it showed a

combination level of education, household size, access to credit and access to training significantly affected adoption. This agrees with findings from other studies whereby, similar socio-economic characteristics of the smallholder farmers affected the adoption of agroforestry (Akpabio and Ibok 2009; Maluki *et al.* 2016; Mukungei *et al.* 2013; Oino and Mugure 2013). Education improves knowledge, awareness, management skills and extension services in agroforestry (Thangata 1996) leading to likelihood of adoption. In the study, agroforestry was adopted better among households with primary and secondary levels of education compared with those having higher educational qualifications. Although this is quite unexpected, it concurs studies within the Sub Saharan Africa (Nyanga *et al.* 2016; Rotich *et al.* 2017; Sood and Mitchell 2009). Although the literacy levels is supposed to determine the levels of agroforestry adoption, in most of the countries within the Sub Saharan Africa region, most people who have basic education appear to be the most active in agricultural adoptions. Majority of the adopters of agroforestry had household size ranging between 6-10 family members to provide labour. The best household size that favoured adoption of agroforestry was large household size with 6-10 people, suggesting that higher adoption rates of agroforestry were related to the availability of family labour. Large household size positively influences adoption of labor-demanding agricultural technologies since they have the ability to relax the labor limitations necessary in the course of introduction of new technologies. Labour from the majority of household members who fall in lower age brackets is restricted because these groups spend most of their time studying in schools and colleges. However, these studies are not in agreement with those of (Uisso and Masao 2016) who did not find any significant relationship between household size and agroforestry adoption and practices. The combinations of these factors have been highlighted by several studies to be crucial in providing the adopters with knowledge, manpower and technical ability to undertake agroforestry practices (Coulibaly *et al.* 2017).

As concern the institutional factors, access to credit facilities and access to formal agroforestry training significantly affected adoption of agroforestry in the study area similar to other studies (Kiptot *et al.* 2006; Mukungei *et al.* 2013). Extension strategies, including field schools, exchange visits and farmer training, are effective ways of disseminating agroforestry information. Unfortunately, agricultural extension officers concentrated on crops and animal production, while on the other hand, Forest Extension officers embarked on tree planting activities only. Many agricultural extension workers are not familiar with trees and shrub species that could fit in an agroforestry system. These agricultural trained extension agents have little knowledge about agroforestry trees with respect to their vernacular names, ecology, propagation, management and uses. On the other hand, forestry extension workers tend to view tree species from a purely "forestry" point of view, and neglect the needs and constraints identified by farmers. Most of the respondents in Kapsaret cited faulty extension services, with inadequate follow up visits or insufficient time for training and advice (Ipara 1993) observed that poor extension services and understaffing were the main bottlenecks to agroforestry technology adoption by women in Vihiga Division in Kenya. Likewise, farmers in Kapsaret believe that there is a direct influence of extension services.

5. CONCLUSION

The study determined that 82.3% of the respondents adopted agroforestry in the form of boundary planting (73.8%), hedgerow (69.4%), scattered planting (51.2%) and alley cropping (37.1%). The study established that level of education, household size, access to credit and formal training were the significant socio-economic and institutional factors affecting adoption of agroforestry. Highest level of adoption of agroforestry occurred among those with primary and secondary levels of education, increased with large household size with 6-10 people and was most adopted by households with access to credit facilities and formal agroforestry training. Based on the findings of study, we recommend: need for capacity building to raise farmers' level of awareness of the agroforestry practices; need for agroforestry extension services for the smallholder farmers on agroforestry adoption. Also the farmers need to be educated on the appropriate agroforestry practices.

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