Technical Efficiency of Sweet Potato Production: A Stochastic Frontier Analysis

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A B S T R A C T

The level of yield among sweet potato farmers is on a decline; low output and yield differences was observed, indicating the existence of inefficiency in production systems and variations in input utilization. Efficiency in resource use must be sustained in order to improve productivity and maximize farm output. This study therefore analyzed the technical efficiency of sweet potato production. Multi-stage sampling techniques were adopted in selecting 94 respondents for this study. Data collected was analyzed using descriptive statistics and stochastic frontier production function. The socioeconomic variables of the respondents affected their farm efficiency and level of farm output. The estimated ratio of the L/R test was 0.579; indicating a goodness of fit of the frontier model and thus a rejection of the null hypothesis. The coefficients of sweet potato seeds (vines) (0.362) and labour (0.439) were positive and statistically significant at 5% level of probability, while the coefficients of farm size (-1.333), fertilizer (-0.452) and herbicides (-0.766) were negative but statistically significant at 5% level of probability. The inefficiency model revealed that the coefficient of farm capital (-0.172), education (-2.281), access to credit (-0.472), farming experience (-0.639), extension contact (-0.733) and membership of cooperatives (-0.396) were negative and statistically significant at 5% level of probability. The mean technical efficiency was 0.62 (62%) implying that the sweet potato farmers in the study area were not producing at optimal capacity. The constraints identified significantly affected sweet potato production in the study area. Subsidizing input costs; sensitizing farmers on appropriate farming practices, cooperative formation and efficiency in resource utilization; improving access to agricultural inputs, technology, farm capital, credit and extension services, market linkages, farm labour supply and the development of indigenous technologies in sweet potato production are strongly recommended.

Introduction

Sweet Potato (Ipomoea Batatas) is a root vegetable that produces tuberous roots (Huntington Lib., 2006). It is a tropical root crop, with more than 100 million tons produced globally per annum (Warammboi et al., 2011; UNCTAD, 2012; FAOSTAT, 2007). Propagation is by adventitious roots, stem, or root cuttings (Huntington Lib., 2006). China currently accounts for more than half of the total global sweet potato output at 55 million metric tons per annum. Nigeria is among the world’s largest producers of sweet potato with an average of over 1 million metric tons produced annually (Huntington Lib., 2006), yet the average yield of 7 tons/ha in Nigeria were below the yield potential of about 35 tons/ha in China using similar labour intensive and technology (FAOSTAT, 2007). However, 15 tons per hectare is attainable by farmers in Nigeria using improved varieties (Okonkwo and Okoli, 2000). In a study carried out by Okonkwo and Okoli (2000) it was revealed that while irrigated irish potato production is an economically viable venture, it is more economically reasonable to cultivate (grow) sweet potato under rain fed condition in the study area: which implies that farmers here monopolize the agro-ecological environment in growing this crop during the rainy season. Despite the importance of sweet potato, it is considered a minor crop in terms of
production and consumption in Nigeria (Adewumi and Adebayo, 2008; UNCTAD, 2012; Woolfe, 2002). In a study carried out by FAOSTAT (2007), it was reported that 115 countries produced 106,569,572 tons of Sweet potatoes in 2010. However, supply remains very concentrated; 82.3% of the global production being in Asia with 81,175,660 tons, China produced by far the largest part and possesses a little less than half of the global acreage dedicated to the sweet potato. Indonesia is the second Asian producing country and the 4th in the world with more than 2 million metric tons of production, for example in the Papua province in Indonesia, 90% of the dishes contain sweet potato. The second continent in the world of sweet potato production is Africa. Africa contributed up to 14% of the global production with more than 14.2 metric tons. Contrary to the main producing countries which have seen their production level decline over the years, some Africa countries have increased their production level from 2 metric ton in 1999 to 2.83 metric ton in 2010 (FAOSTAT, 2007). Nigeria and Tanzania produce 1.43Mt and 1.4Mt respectively (FAOSTAT, 2007). This expansion in Africa is linked to a strong demographic growth. However, Latin America which is the global sweet potato arena, produced 1.97mt in 2010, that is, a little more than 2% of the global supplies. In a study carried out by FAOSTAT (2007) it was reported that Nigeria was ranked the second largest producer of sweet potato after Uganda in Africa. In a study carried out by UNCTAD (2012) it was reported that sweet potato production in terms of land size used from 2002 to 2012 depicts a law of diminishing returns. In this report, from 2003 to 2006, there were increased and moderate relationships between production per tones and land sizes used, but in 2007 there was a sharp decrease in production even though the land size was increased to about 10.8%. This increase was the biggest in the whole decade while the 2008 gave a remarkable change. Production figure rose to about 36.4% and land size declined to about 2.2%. But the periods 2009, 2010 and 2011 show that production and land size used remained unchanged with yields figures of 15000 tons/ha. Lastly, in 2012 the yield figure was increased to 1.6%. The cultivation and utilization of sweet potato have not received appropriate attention of the Nigerian populace despite its nutritional constituents, ease of propagation, soil conservation attribute and industrial use (Woolfe, 2002). It was regarded as a crop with little economic importance. Its consumption was surrounded by the erroneous idea that it caused amoebic dysentery (Woolfe, 2002). The minimal utilization of sweet potato in Nigeria may also be attributable to non-availability of adequate sweet potato- based recipes that satisfy the food habits of Nigerians (Adewumi and Adebayo, 2008; Warammboi et al., 2011). Of the estimated 150 million tons of all root and tubers produced in Nigeria annually, sweet potato contributes only 13% (Horton, 2008; Ekwelle et al., 2001). In Nigeria, the production, marketing and utilization of sweet potato have expanded to almost all the ecological zones within the past decade (Ekwelle et al., 2001; Adu-Kwarteng et al., 2002; FAOSTAT, 2007). Sweet potato is the second highest source of energy crop after cassava producing 465KJ, that is, only 125KJ less than cassava (SPU, 2013). Industrially, sweet potatoes flour can be used to substitute wheat flour in bread making or maize flour in balanced feeds. Energy is measured in kilocalories (Kcal), calories or Kilojoules (KJ), and 1 kilocalorie = 4.2 kilojoules). However, industrial potentials of sweet potato have not been exploited due mainly to a chronic lack of awareness about the numerous commercial benefits derived from it (Azogu and Olomo, 2002).

The analysis of efficiency is generally associated with the possibility of farms producing a certain optimal level of output from a given bundle of resources or a certain level of output at least cost (Amaza, 2007). Efficiency can be defined as the relative performance of the processes used in transforming input into output (Amaza, 2007). It could also be defined as the attainment of production goals without waste (Ajibefun et al., 2002). The pivotal role of efficiency in accelerating agricultural productivity and output has been applauded and investigated by numerous researchers within Africa and outside. The decreased output of food crop production over the years may not only be connected with deviations of farmers’ practices from technical recommendations but also with the use of resources at sub-optimal levels which ultimately leads to technical and allocate inefficiencies (Coelli et al., 1998). An underlying premise behind much of the research inefficiency is that farmers are not making efficient use of existing technology, then efforts designed to improve efficiency would be more cost-effective than introducing new technologies as a means of increasing agricultural output (Adeleke et al., 2008). Broadly, two quantitative approaches are developed for measurement of production efficiency: Parametric (Stochastic frontier approach) and non-parametric (Data envelopment analysis) approaches. The data envelopment analysis (DEA) has no fixed functional form and does not account for noise in the data. Thus, all deviation from the frontier will be accounted for as inefficiencies (Amaza, 2007). The Stochastic frontier approach is parametric and is sensitive to the choice of functional form and accounts for random errors. In this approach all deviations from the frontier are due to random effects and inefficiency (Coelli et al., 2002). The stochastic frontier production was independently proposed by Aigner et al. (1977) and Mees were and Van den Broeck (1977). The measurement of efficiency is important because it leads to substantial resource savings. Early application of stochastic frontier production functions was first used in the analysis of the United States agricultural data. Battese and Corra (1977) applied the techniques to the pastoral zone of Eastern Australia. And more recently, empirical application of the techniques in efficiency analysis have been reported (Ojo and Ajibefun, 2000; Ojo, 2003; Dang, 2006; Usman, 2009; Amaza, 2007). Several studies have identified numerous socioeconomic variables that influence efficiency of inputs use. These factors include age, education, farmer’s experience, contact with extension agents, income of the farmers and access to credit. Awudu and Richard (2001) reported that efficiency increased with the age until a maximum efficiency was reached. Alene and Hassan (2003) reported that technical efficiency of Ethiopian farmers was positive and significantly influenced by education level, credit and contact with extension workers. Ugunyinka and Ajibefun (2003) observed that education and membership of Farm association were the most important factors affecting efficiency. Education level and farming experience have
been reported to have a positive and significant impact on technical efficiency (Adewuyi and Okumadewa, 2001). Extension contact has been reported to have a positive and significant relationship with efficiency (Amaza, 2007). Therefore, farmers that have had extension contacts are likely to be more efficient than those without any extension contact. Greater family size increase efficiency (Belen et al., 2003). This can be explained by the fact that readily available family labour will allow for the timely execution of important farm activities such as fertilization and weeding, thus, contributing to higher yields. Besides, most farmers are financially constrained and thus, the availabilities of family labor will ease hiring of labour. Farm size has been reported to have a positive and significant relationship with technical efficiency (Tewe et al., 2003). One serious problem facing Nigeria today is chronic and transitory food insecurity (World Bank, 2003). Sweet potato is highly regarded as a food security crop and it is the most productive crop among all the other staple crops and tolerates occasional dry spells and yields even on less fertile soil in contrast to other crops such as maize (Woolfe, 2002; Zuraida, 2003). In a study carried out by World Bank (2003) it opined that, despite the fact that Nigeria was found to be the second highest producer of sweet potato in Africa, it was ranked 17th in terms of output produced per land area, suggesting that sweet potato producers in Nigeria are quite inefficient in relation to farmers in other African countries. There is great need to improve the national production from over 1 million to 5 million tons per annum (World Bank, 2003). In Nigeria the output from sweet potato production is low and therefore there is need to empirically investigate factors that affect farm efficiency and productivity. This research determines factors of farm efficiency that can boost the level of farm productivity. It would add to the existing body of knowledge in rural sociology and economics of root crop production. It will also provide policy makers, development planners and other stakeholders with necessary data and insight for effective and sustainable policies and programmes that would facilitate and boost farm productivity and efficiency. Therefore, this study analyzes the technical efficiency of sweet potato production, while attempting to address the following research questions:

- What are the socioeconomic characteristics of the respondents?
- What is the technical efficiency of sweet potato production?
- What is the efficiency index among sweet potato farmers?
- What are the constraints of sweet potato production?

H₀: There is no significant relationship between inputs and output in sweet potato production.

**Material and Methods**

**Study Area**

This study was carried out in Bokkos Local Government Area (LGA) of Plateau State, Nigeria. It has a total area of 1682 km² and located between latitude 9°15' N and 8°3' E, with a total projected population of 392,026 in 2016 (FAOSTAT, 2007). The Local Government is made up of 8 districts which include; Mushere, Daffo, Richa, Sha, Manguna, Toff, Kamoi and Bokkos. The LGA is located at the central region of Plateau State and it is surrounded by rocks and scattered vegetation. Its Annual rainfall averages from 600mm-1000mm, with average temperatures of about 24°C-29°C annually (FAOSTAT, 2007). The major crops cultivated in the study area are Irish potato, sweet potato, cocoyam, maize and red beans.

**Sampling Technique**

Multistage sampling technique was used in selecting the respondents for the study. The first stage involved the Purposive selection of Bokkos LGA due to the prevalence of sweet potato production in the study area. The second stage involved the selection of three (3) districts out of eight (8) in the study area due to the prevalence of sweet potato farmers in the selected districts (Daffo, Sha and Bokkos districts). The third stage involved the collection of a compiled list of sweet potato farmers from Plateau state ADP extension agent at the LGA secretariat. In the last stage, using the list of estimated population of sweet potato farmers in the selected districts, respondents were randomly selected using 0.2 sampling proportion. Based on the foregoing, 94 respondents were randomly selected for the study. Table 1 presents the sample frame distribution.

| Selected Districts | Communities | Sample Frame | Sample Size |
|--------------------|-------------|--------------|-------------|
| Daffo              | Ganda       | 135          | 27          |
|                    | Magi        | 64           | 13          |
|                    | Ngajul      | 37           | 7           |
|                    | Kumnet      | 66           | 13          |
| Bokkos             | Mangar      | 35           | 7           |
|                    | Tarangol    | 34           | 7           |
|                    | Manguna     | 61           | 12          |
| Sha                | Tar         | 38           | 8           |
| Total              |             | 470          | 94          |

Source: Plateau State ADP, 2017

**Validation of the Research Instrument**

Content validity was used to measure the adequacy of the instrument items in this study. Content validity in this context sought to determine the relevance and adequacy of items included in the instruments. Using the Jury Method (Kerlinger, 1973), the entire instrument was subjected to the scrutiny of relevant experts. Each of the experts was requested to independently give his expert opinion on the relevance and adequacy of the items with respect to the objectives of the study. Various questions of the data collection instrument were scrutinized in terms of how relevant they are to the specific objectives of the study as well as how the prepared questions exhaustively cover the specific objectives of the study. Furthermore, the data collection instrument was examined against the background of its adequacy in regard to the accomplishment of the objectives of the study.
**Instrument Reliability Test**

An instrument is considered reliable when it consistently produces the same result when applied to the same sample many times (Osuala, 2005). The test-retest method of affirming instrument reliability was employed for this study. It was computed by calculating the correlation coefficient between two distributions of test scores obtained at two different times on the same respondents. The instrument was trial tested on 20 respondents drawn from two districts in the Local Government Area viz: Daffo and Bokkos. The information obtained from the responses to the instrument were analysed using product-moment correlation analysis. High value of mean product-moment correlation coefficient of 0.735 indicated high reliability of the instrument.

**Method of Data Collection**

Primary data were collected from the sweet potato farmers in the study area, with the use of structured questionnaires in line with the specific objectives of the study.

**Analytical Techniques**

Descriptive statistics (such as frequency distribution, percentages and mean) was used to analyze objectives i and iv, while the stochastic frontier production model was used to analyze objective ii and iii.

**Stochastic Frontier Production**

The Stochastic frontier approach is parametric and is sensitive to the choice of functional form and accounts for random errors. In this approach all deviations from the frontier are due to random effects and inefficiency. Efficiency in resource use must be sustained in order to improve productivity and maximize farm output. Technically efficient production is defined as the maximum quantity of output attainable by a given input (Pitt and Lee, 1981). According to Njeru (2004); technical efficiency is the ability of a firm to maximize output for a given set of resource inputs. Farm efficiency and productivity are indicators of agricultural sustainability (Goni et al., 2013) and a prerequisite for optimum farm production since inefficiency in resource use can distort food availability and security (Etim et al., 2005). Efficiency measurement is germane in production studies. Inefficiency in the use of available scarce resources has been the bane of increased food production. According to Njeru (2004), technical efficiency is the ability of a firm to maximize output for a given set of resource inputs. Agricultural economists always provide the guidance to farmers about efficient utilization of inputs. Efficient utilization of inputs is also important for food security (Irz et al., 2010). The modeling and estimation of stochastic frontier production functions are useful to provide information about the relationship between the amount of output and the inputs of production, given the level of technology involved. In recent years, stochastic frontier models in agricultural economics have been used. The stochastic frontier model was originally proposed for the analysis of the panel data by Battese and Coelli (1995). The stochastic frontier production model is estimated using the maximum likelihood estimation procedure (MLE) (Battese and Corra, 1977). The technical efficiency of an individual firm is defined in terms of the observed output (Yi) to the corresponding frontier output (Yi*) given the available technology as specified in equation (1) and (2);

\[
\text{TE}_i = \frac{Y_i}{Y_i^*} \quad (1)
\]

\[
\text{TE}_i = f(x_i; \beta) \exp (v_i - u_i) / f(x_i; \beta) \exp(v_i) - \exp (-u_i) \quad (2)
\]

So that \(0 \leq \text{TE}_i \leq 1\)

Therefore, the technical inefficiency is equal to 1-TE. However, a general stochastic frontier production function for the cross-sectional data, which is considered in this paper, is defined implicitly in equation (3);

\[
Y_i = \beta_i X_i + V_i - U_i \quad (3)
\]

Where;

\(Y_i\) = denotes the output for the \(i^{th}\) sample farm;
\(\beta_i\) = vector of unknown parameters to be estimated;
\(X_i\) = vector of explanatory variables for the \(i^{th}\) farm;
\(V_i\) = independent and identically distributed random errors which have normal distribution with unknown variance \(\alpha^2\);
\(U_i\) = non-negative unobservable random variables associated with the technical inefficiency of production, such that for a given technology and levels of inputs, the observed output falls short of its potential output.

Technically inefficiency effect model proposed in a study as adapted from Battese and Coelli (1995) is implicitly presented in equation (4);

\[
U_i = \delta_0 + \delta_1 Z_i \quad (4)
\]

Where;

\(U_i\) = Technical inefficiency
\(\delta_0\) = vector of unknown parameters;
\(\delta_1\) = vector of parameters to be estimated; and
\(Z_i\) = explanatory variables associated with the technical inefficiency effects.

The stochastic frontier production function model was employed to analyze objective (ii). The stochastic production function with a multiplicative disturbance term is presented in equation (5);

\[
Y = f(X\beta) + e_i \quad (5)
\]

Where \(Y\) is the farm output in kg, \(X\) is a vector of input quantities; \(\beta\) is a vector of parameters and \(e\) is a stochastic disturbance term consisting of two independent elements \(U\) and \(V\), given by:

\[
e_i = v - u \quad (6)
\]

The empirical model stochastic frontier production function used in this study is specified in a double log form of Cobb–Douglas production function. The Cobb Douglas
function is very useful in empirical analysis. The partial elasticity’s are equal to each of the parameters and when linearized in log, the function is easy to fit and the coefficients are direct elasticity’s. The Cobb-Douglas production function is specified in equation (7);

\[ \ln Y_i = \beta_0 + \beta_1 \ln X_1 + \beta_2 \ln X_2 + \beta_3 \ln X_3 + \beta_4 \ln X_4 + \beta_5 \ln X_5 + \ldots + u_i \]  \hspace{1cm} (7)

Where;
\[ \ln = \text{natural logarithm to base } e, \]
\[ Y_i = \text{output of sweet potato (kg/ha)} \]
\[ V_i = \text{random error due to mis-specification of the model and variation in output due to exogenous factors outside the farmer’s control.} \]
\[ U_i = \text{technical inefficiency effects which captures deviation from the frontier.} \]
\[ \beta_0 = \text{intercept} \]
\[ \beta_1, \ldots, \beta_n = \text{Unknown parameters which are to be determined.} \]
\[ X_1 = \text{farm size measured in hectares (ha)}; \]
\[ X_2 = \text{quantity of seed used (number of vine cuttings/ha)}; \]
\[ X_3 = \text{labour measured in man-days;} \]
\[ X_4 = \text{quantity of fertilizer measured in kilogram per hectare (kg/ha)}; \]
\[ X_5 = \text{quantity of herbicides applied measured in litres per hectare (litre/ha)} \]

**Technical Inefficiency Effect Model**

The average level of technical inefficiency measured by the truncated normal distribution has been assumed to be a function of socioeconomic factors and internal transaction costs. In the inefficiency effect model, a positive sign of an estimated parameter implies that the associated variable has a negative effect on efficiency but positive sign of an estimated parameter implies that the technical inefficiency and its determinants in crop production specified in equation (8);

\[ U_i = \alpha_0 + \alpha_1 Z_1 + \alpha_2 Z_2 + \alpha_3 Z_3 + \alpha_4 Z_4 + \alpha_5 Z_5 + \alpha_6 Z_6 + \alpha_7 Z_7 + \ldots + w_i \]  \hspace{1cm} (8)

Where;
\[ U_i = \text{technical inefficiency effects of the ith farmer;} \]
\[ \alpha_0 = \text{intercept} \]
\[ \alpha_1, \ldots, \alpha_n = \text{parameters to be estimated;} \]

Where:
\[ Z_1 = \text{Farm capital measured in N (naira)}; \]
\[ Z_2 = \text{Education measured in years;} \]
\[ Z_3 = \text{Access to formal credit (Yes = 1, No = 0)}; \]
\[ Z_4 = \text{Farm experience measured in years;} \]
\[ Z_5 = \text{Extension contact (Yes = 1, No = 0)}; \]
\[ Z_6 = \text{Membership of cooperatives/farmers association (Yes=1, No=0)} \]
\[ Z_7 = \text{Household size (population)} \]
\[ w_i = \text{is the random variable which is defined by the truncation of the normal distribution with zero mean and variance.} \]
\[ Z_i = \text{is expected to be negatively related to the level of inefficiency in sweet potato production} \]

**Results and Discussion**

**Socioeconomic Characteristics of Sweet Potato Farmers**

**Age of the Respondents**

Table 2 revealed that most (56.4%) of the respondents were within the age bracket of 239 years; suggesting that most of the respondents were in their economically productive age bracket. The mean age of the respondents was 4.1 years. This result implies a great prospective for sustainable sweet potato production in the study area. The result in Table 2 reveals that the farmers were strong, agile and active and can participate adequately in farming activities. Age is expected to have negative influence on the respondent’s participation in improved sweet potato production that is why younger farmers are more active in the production of this crop. This result agrees with the findings in a study carried out by Amaza, 2007 and Akoneda (2009), who also reported similar results of the age of farmers engaged in agricultural production.

**Gender of the Respondents**

Table 3 revealed that most (57.4%) of the respondents in the study area were male, while 42.6% are females; this indicates that the respondents were predominantly men; hence gender was also an essential socioeconomic factor that influences access to farm assets and resources agricultural production in Nigeria. This result agrees with the findings in a study carried out by Amaza, 2007 on farmer’s demography.

**Farm size**

Table 4 revealed that most (63.8%) of the farmers had farm holdings of 219.9 ha, 25.5% of had farm holdings of 20.39 ha, while 10.7% constitutes farmers with 40.0 ha. The mean farm size was 1.3ha, implying that most of the farmers in the study area were producing at subsistence level and the likely implication of this is low farm output. This result agrees with the findings in a study carried out by Amaza, 2007 and Belon et al., 2003 on farmer’s demography.

**Table 2. Distribution based on the Age of the Respondents**

| Age (years) | Frequency | Percentage (%) |
|-------------|-----------|----------------|
| ≤39         | 53        | 56.4           |
| 40-59       | 31        | 33             |
| >60         | 10        | 10.6           |
| Mean age = 41.4 |        |                |

Source: Field survey, 2017

**Table 3. Distribution based on the Gender of the Respondents**

| Gender | Frequency | Percentage (%) |
|--------|-----------|----------------|
| Male   | 54        | 57.4           |
| Female | 40        | 42.6           |

Source: field survey 2017

**Table 4. Distribution of the respondents based on their farm size**

| Farm size (ha) | Frequency | Percentage (%) |
|----------------|-----------|----------------|
| ≤1.9           | 60        | 63.8           |
| 2.0-3.9        | 24        | 25.5           |
| ≥4.0           | 10        | 10.7           |
| Mean = 1.3ha   |           |                |

Source: field survey 2017
Household Size of the Respondents

Table 5 revealed that most (59.6%) of the respondents had a household population of ≤9 people. The respondents have a mean household size of 7 people per household. The larger the size of the household the more labour supply for farm activities. This implies that households in the study area had adequate labour supply to embark on expansionary farming activities which will result to increased farm efficiency and productivity at low labour cost. This result agrees with the findings in a study carried out by Amaza, 2007 and Coelli et al., 2002 on farmer’s demography.

Farm Labour Supply

Table 6 revealed that most (60.6%) of the respondents used both family and hired labour in carrying out various operations of their farm activity; 21.3% used mainly family labour while hired labour was (18.1%) for their sweet potato farming activities. For those that supplement the hired labour with family labour, family labour was used in operations such as planting, fertilizer application and harvesting while hired labour was used mainly for land clearing, ridging, weeding and spraying. A majority of the respondents claimed to have experienced shortage of labour during land clearing and ridging leading to a high cost of performing such operations. The average labour input per household is 124 man-day ha⁻¹, out of which 52% was family labour. This result agrees with the findings in a study carried out by Amaza, 2007 and Cechura et al., 2014 on farmers demography.

Farming System

Table 7 revealed that most (78.7%) of the respondents in the study area were subsistent farmers, while 21.3% were commercial farms which were mostly communal farms. This predominant farm system was attributable to the prevalent tenure policies which caused fragmentation of most potential farms lands; resulting to a prevalence of small farm holdings among most farmers in the study area. This result agrees with the findings in a study carried out by Amaza, 2007 on farmer’s demography.

Management Practice

Table 8 revealed that most (53.2%) of the respondents in the study area adopted mixed cropping systems, while 46.8% adopted monocropping systems. This predominant cropping system is attributable to the farm size of the respondents; hence most farms combined cultivation of sweet potato production with other crops so as to maximize a variety of farm output relative to their small farm holdings in the study area. This result agrees with the findings in a study carried out by Amaza, 2007 and Etim et al., 2005 on farmer’s demography.

Farm Capital

Table 9 revealed that most (71.3%) of the respondents in the study area used their personal savings to finance their farm activities. Capital from this source is usually very small and this may be one of the reasons why the farmers cultivated at very subsistent level with inadequate capacities to scale-up their farm activities, while 28.7% got credit from other sources to supplement their personal savings. Agricultural credit enables farmers to augment their farm capital. The mean farm capital per respondents was N71,500. This result agrees with the findings in a study carried out by Amaza, 2007 on farmers demography.

Access to Credit

Table 10 revealed that most (81.9%) of the respondents in the study area do not have access to agricultural credit. Agricultural credit helps farmers to augment their farm capital; however, respondents in the study were excluded from financial services due to the absence of financial institutions in the study area. Adequate funding is required by farmers to finance all sweet potato production activities. However, a large number of farmers face serious shortage of funds to finance their sweet potato production activities, which in turn limits their level of production. This low access to credit could also be attributed to the fact that government seldom grants financial credit to large number of farmers. In a study carried out by Ekong (2003) it was asserted that credit is a very strong factor that is needed to acquire or develop any enterprise; its availability could determine the extent of production capacity.
Membership of Cooperative of Respondents

Table 11 revealed that most (84%) of the respondents do not belong to any cooperative society. This affects the ability to boost levels of farm efficiency and output through synergy in agricultural resource utilization among farmers. Also, membership of farm associations or cooperative societies avail farmers the opportunity to have more access to agricultural credit, receive agricultural inputs at subsidized rates and for effective information dissemination on improved agricultural practices and technology that boost levels of farm efficiency and output. Consequently, farmers who belong to the cooperative societies enjoy the benefits accruable to members through the pooling of resources together for better expansion of their production frontier; efficient and effective management of resources and for profit maximization. In a study carried out by Wilson et al. (2001) and Ekong (2003) it was reported that membership of cooperative societies has advantages of accessibility to micro-credit, input subsidy and also as avenue in cross breeding ideas and information.

Farm Output of the Respondents

Table 12 revealed that most (80.9%) of the respondents obtained a farm output of $\leq 4.9$ tons per hectare. The respondents had a mean farm output of $3.1$tons ha$^{-1}$. This low level of farm output can be attributable to their farm size as well poor resource utilization among the farmers which can maximize farm output. This implies that majority of the farmer where producing at subsistent level in the study area. This result agrees with the findings in a study carried out by Amaza, 2007 and Alene and Hassan, 2003 on farmer’s demography.

Farming Experience of the Respondents

Table 14 revealed that most (44.7%) of the respondents attained primary education. This implies that majority of the respondents were literates and had basic educational backgrounds, which is in turn could enhance their productive capacities. Research has shown that education enables farm households in rural areas to adopt new agricultural methods, cope with risk, respond to market signals and improve agricultural productivity (Njeru, 2004).

Table 11. Distribution based on Membership of Cooperatives

| Membership | Frequency | Percentage (%) |
|------------|-----------|----------------|
| Yes        | 15        | 16             |
| No         | 79        | 84             |

Source: field survey, 2017

Table 12. Distribution based on Farm Output in Tons per Hectare (Tons ha$^{-1}$)

| Output level (Tons ha$^{-1}$) | Frequency | Percentage (%) |
|------------------------------|-----------|----------------|
| $\leq 4.9$                   | 76        | 80.9           |
| 5-9.9                        | 13        | 13.8           |
| $\geq 10$                    | 5         | 5.3            |
| Mean = 3.1                  |           |                |

Source: field survey 2017

Table 13. Distribution based on quantity of seed (vine) used per Hectare (gram ha$^{-1}$)

| Seed quantity (gram ha$^{-1}$) | Frequency | Percentage (%) |
|--------------------------------|-----------|----------------|
| $\leq 999$                     | 75        | 79.8           |
| 1000-2999                      | 14        | 14.9           |
| $\geq 3000$                    | 5         | 5.3            |
| Mean = 700 grams               |           |                |

Source: field survey 2017

Table 14. Distribution of the respondents based on their educational level

| Educational level | Frequency | Percentage (%) |
|-------------------|-----------|----------------|
| Non-formal        | 8         | 8.5            |
| Primary           | 42        | 44.7           |
| Secondary         | 29        | 30.9           |
| Tertiary          | 15        | 16             |

Source: field survey 2017

Table 15. Distribution based on the Farming Experience of the Respondents

| Farming experience | Frequency | Percentage (%) |
|--------------------|-----------|----------------|
| $\leq 9$           | 23        | 24.5           |
| 10-19              | 51        | 54.3           |
| $\geq 20$          | 20        | 21.3           |
| Mean = 18 years    |           |                |

Source: field survey 2017

Potato production. The result opined that the respondents had adequate experience necessary for increased production. This shows that the managerial ability of farmers can be inferred to be reasonably good. This is because the more experienced a farmer is the more efficient his decision making processes and willingness to take risks associated with adoption of innovation to increase farm productivity. Farming experience is the act of gaining knowledge through constant practicing of skill, which brings about specialization (Belon et al., 2003). Experienced farmers have the ability to use modern agricultural technology and practices. In a study carried out by Tewe et al. (2003) it was reported that experience enhances more efficient use of scarce resources by small-scale farmers in his study of socioeconomic determinants of output and profit levels of small-holder rice production systems in Abia State.
Agrochemical Application

Table 16 revealed that most (83%) of the respondents in the study area applied herbicides on their farms, while 17% used pesticides on their farms. An average of 10 liters ha^{-1} of herbicides was applied by the respondents on their sweet potato farms. This results indicates that majority of the respondents applied herbicides on their farms for the purpose of weed management and control in the study area. Agrochemicals are expensive and not readily available in the study area; hence they need for more extension activities to sensitive farmers on the use of these agrochemicals and subsidization of agricultural input prices is also required. This result corroborates with the findings of Ajetomba (2005) and Olarinde et al. (2005) who also reported similar results on agrochemical application in arable crop production.

Fertilizer Application

Table 17 revealed that most (89.4%) of the respondents in the study area applied organic fertilizers on their farms, while 10.6% used inorganic fertilizers on their farms. An average of 250 kilogram (kg) ha^{-1} of inorganic fertilizer was applied by the respondents on their sweet potato farms. This results indicates that majority of the respondents applied organic fertilizers on their farms; suggesting a prevalence of organic farming activities among respondents in the study area. Application of organic fertilizers can be attributable to high cost of procurement and non-availability of inorganic fertilizers as compared to organic fertilizers, which were relatively cheaper and more accessible to the farmers. However, the use of fertilizer among the respondents was grossly inadequate, some respondents also claimed that they have fertile land that does not require fertilizer application. This result corroborates with the findings in a study carried out by Ajetomba (2005) and Olarinde et al. (2005) who reported similar results on fertilizer application in arable crop production.

Extension Contact

Table 18 revealed that most (87%) of the respondents in the study area had no access to extension. This indicates that they respondents do not have access to agricultural information and technology that will boost their level of farm efficiency and output. Based on their literacy levels the respondents would have easily adopted improved agricultural technology. Extension contact is germane to build farmers capacity, resilience to agricultural risks, poverty reduction strategies, innovation and practices that will ensure sustained farm efficiency and food security in the study area. This result agrees with the findings in a study carried out by Belon et al. (2003) on farmer’s demography.

Technical Efficiency of Sweet Potato Production

The estimates of the parameters of the stochastic frontier model are presented in Table 18. The estimated value of the coefficient of Sigma square (\sigma^2) (0.699) was positive and significantly different from zero at 5% (P<0.05) level of probability. This indicates a goodness of fit of the model and correctness of the specified distributional assumptions of the composite error terms for the regression analysis. The value of gamma (\gamma) is estimated to be 27% and not statistically significant. This is consistent with the theory that true \gamma-value should be greater than zero. This implies that 27% of random variation in the yield of the farmers was due to the farmers’ inefficiency on their respective farms and not as a result of random variability. Since these factors are under the control of the farmer, reducing the influence of the effect of \gamma will greatly enhance the technical efficiency of the farmers and improve their yield. The gamma \gamma indicates the systematic influences that are unexplained by the production function and the dominant sources of random error. This means that the inefficiency effects make significant contribution to the technical inefficiencies of sweet potato farmers; hence the technical inefficiency effects are significant in the estimated model. The study revealed that the generalized log likelihood function was -88.751. The log likelihood function implies that inefficiency exist in the data set. The estimated ratio of the L/R test was 0.579. The L/R test assessed the goodness of fit of the statistical models based on the ratio of their likelihoods; it compared the deterministic production and stochastic function in selecting the appropriate functional form (cobb-Douglas); hence the estimated ratio was significantly different from one and equivalently its natural logarithm was significantly different from zero (0<1); indicating a goodness of fit of the frontier model. Thus the null hypothesis was rejected. The average technical efficiency for the farmers is 0.62 implying that, on the average the respondents were able to obtain 62% of potential output from a given mixture of production inputs. Thus, in a short run, there is minimal score of increasing the efficiency by 38%, through adoption of modern technology and techniques in sweet potato production.

Table 16: Distribution based on Agrochemicals Applied by the Respondents

| Agrochemical   | Frequency | Percentage (%) |
|----------------|-----------|----------------|
| Herbicide      | 78        | 83             |
| PM             | 16        | 17             |

PM: Pesticide Mean quantity 10 liters ha^{-1}, Source: field survey 2017

Table 17: Distribution based on Fertilizer Applied by the Respondents

| Agrochemical   | Frequency | Percentage (%) |
|----------------|-----------|----------------|
| Organic        | 84        | 89.4           |
| IM             | 10        | 10.6           |

IM: Inorganic Mean quantity 250 kg ha^{-1}, Source: field survey 2017

Table 18: Distribution based on the Extension Contact of the Respondents

| Extension contact | Frequency | Percentage (%) |
|-------------------|-----------|----------------|
| No                | 74        | 78.7           |
| Yes               | 20        | 21.3           |
| Total             | 92        | 100            |

Source: field survey, 2017

In consonance with a priori expectation, the coefficients of sweet potato seeds (vines) (0.362) and labour (0.439) were positive and statistically significant at 5% (P<0.05) level of probability; implying that an increase in these variables will increase the level of output. This result is in line with the studies carried out by carried out by Okonkwo and Okoli (2000), Amaza (2007) and Coelli et al., (2002), who reported positive correlation in input-
output relationships in crop production. However, against a priori expectation, the coefficients of farm size (-1.333); was negative but statistically significant.

**Farm size:** the coefficients of farm size (-1.333), was negative but statistically significant at 5% (P<0.05) level of significance; suggesting an inverse relationship with farm output. This implies that an increase in farm size no positive impact on the output of sweet potato in the study area. In a study carried out by Coelli et al. (2002), it was reported that smaller farms were economically more efficient than larger farms. If farm size is small, farmers are able to combine their resources better (Ogundari and Ojo, 2006).

**Seed:** The estimated coefficient for sweet potato seed (vines) was 0.362 which is positive and statistically significant at 5% level. This implies that a 1% increase of this variable will increase sweet potato output by 0.36%. The significance of seed quantity is however, due to the fact that seed determines to a large extent the output obtained. If correct seed rates and quality seeds are not used, output will be low even if other inputs are in abundance. This result is in line with the studies carried out by (Etim et al., 2005; Ben- Belhassen, 2002; Belon et al., 2003) who observed that the estimated coefficient of seed and labour inputs were positive and significant at 1% level implying that the more seed is applied and the more labour employed the better the output of sweet potato.

**Labour:** The coefficient of labour was 0.439 which is positive and statistically significant at 5% level of probability. This shows the importance of labour in sweet potato farming in the study area. The implication is that a 1% increase in labour supply may increase the output of sweet potato by 0.44%. This is in line with several studies carried out by Amaza (2007) and Coelli et al. (2002) which revealed the importance of labour in farming, particularly in developing countries where mechanization is rare on small scale farms. In the study area, farm labour plays a crucial role in virtually all farming activities. This situation has variably been attributed to the practice of split-plot cropping on small scattered land holdings and lack of affordable equipment (Belon et al., 2003). This result is in line with the studies carried out by (Onyenweaku et al., 2004; Okonkwo and Okoli, 2000; Belon et al., 2003; Allen and Hassan, 2003) who observed that the estimated coefficient of seed and labour inputs were positive as expected and significant at 1% level implying that the more seed is applied and the more labour employed the better the output of sweet potato.

However, against a priori expectation, the coefficients of fertilizer (-0.452) and herbicides (-0.766) were negative but statistically significant at 5% (P<0.05) level of significance; suggesting an inverse relationship with farm output which is attributable to inadequate and improper application of fertilizer and herbicides by the respondents on their farms.

Most of the coefficients in the inefficiency model were negative. Generally, a negative sign on a parameter means that the variable reduces technical inefficiency, while a positive sign increases technical inefficiency. The inefficiency model revealed that the coefficient of farm capital (-0.172), education (-2.281), access to credit (-0.472), farming experience (-0.639), extension contact (-0.733) and membership of cooperatives (-0.396) were negative and statistically significant at 5% (P<0.05) level of probability. This implies that increase in these variables may have the tendency of reducing the inefficiency level in sweet potato production among farmers in the study area.

Table 19. Stochastic Frontier Analysis of efficiency in Sweet Potato Production

| Variable                  | Parameter | Coefficient | Standard error | T-ratio |
|---------------------------|-----------|-------------|----------------|---------|
| **Efficiency model**      |           |             |                |         |
| Constant                  | β₀        | 2.721**     | 0.983          | 2.768   |
| Farm size (x₁)            | β₁        | 1.333**     | 0.472          | 2.824   |
| Seed (x₂)                 | β₂        | 0.362**     | 0.202          | 1.792   |
| Labour (x₃)               | β₃        | 0.439**     | 0.164          | 2.676   |
| Fertilizer (x₄)           | β₄        | -0.452**    | 0.139          | -3.251  |
| Herbicides (x₅)           | β₅        | -0.766**    | 0.681          | 1.124   |
| **Inefficiency model**    |           |             |                |         |
| Constant                  | α₀        | 2.273**     | 0.891          | 2.551   |
| Capital (z₁)              | α₁        | -0.172***   | 0.055          | -3.127  |
| Education (z₂)            | α₂        | -2.281***   | 0.723          | -3.155  |
| Experience (z₃)           | α₃        | -0.639**    | 0.218          | -2.931  |
| Credit (z₄)               | α₄        | -0.472**    | 0.159          | -2.969  |
| Extension (z₅)            | α₅        | -0.733**    | 0.289          | -2.536  |
| Cooperative (z₆)          | α₆        | -0.396**    | 0.158          | -2.51   |
| Household size (z₇)       | α₇        | 0.218**     | 0.161          | 1.354   |
| **Diagnostic statistic**  |           |             |                |         |
| Sigma –square (ơ²)        |           | 0.699**     | 0.251          | 2.784   |
| Gamma (γ)                 |           | 0.271       | 0.282          | 0.961   |
| Log likelihood function   | (log)     | -88.751     |                |         |
| LR test                   |           | 0.579       |                |         |
| Number of observations    |           | 94          |                |         |
| Mean efficiency           |           | 0.62        |                |         |

Source: field survey 2017; ** = 5% (P<0.05), *** = 1% (P<0.01)
Farm Capital

The coefficient of farm capital (-0.172) was negative but statistically significant at 5% (P<0.05) level of probability, suggesting that a 1% increase in farm capital will increase sweet potato yield by 0.17%, suggesting that farm capital increases efficiency in sweet potato production. The negative sign indicates an inverse relationship with technical inefficiency. This implies that farmers with improved farm capital tend to be more efficient in their farm operations through increased capacity to acquire and adopt improved agricultural technology and inputs that will boost the level of their farm output. This result is consistent with the studies carried out by (Njeru, 2004; Battese and Corra, 1977; Belon et al., 2003; Shrestha et al., 2015) who reported that improved farm efficiency can be attributable to increase in farm capital.

Educational Level

The estimated coefficient of educational level (-2.281) was negative and statistically significant at P<0.05 (5%) level of probability. The number of years spent in school is a proxy of the literacy level of the farmers. The results revealed an inverse relationship to technical inefficiency in sweet potato production. This implies that farmers with better education were technically more efficient; hence, literacy increases farmer’s capacity to adopt and efficiently utilize agricultural technology and information that tend to boost farm output. Increased level of education results in better evaluation of management systems, farm decision making and efficient input utilization. This result corroborates with the studies carried out by (Cechura et al., 2014; Awudu and Richard, 2001; Belon et al., 2003), who also reported that improved farm efficiency can be attributable to increase in the level of education. The positive correlation between education and technical efficiency is consistent with previous studies carried out by (Amaza and Maurice, 2005; Coelli et al., 2002; Amaza, 2007).

Farming Experience

The estimated coefficient of farming experience (-0.639) was negative and statistically significant at P<0.05 (5%) level of probability, suggesting that years of farming experience was positively correlated with farm efficiency; this indicates that an increase in the number of years in sweet potato production decreases technical inefficiency. An increase in efficiency may also be attributable to the experience they have gained over time especially with regard to production techniques and combination of resources. This result corroborates with the studies carried out by Amaza, (2007) who also reported a positive correlation between farming experience and farm output. Also, this is in line with the study carried out by Ogundari and Ojo (2006), and Awudu and Richard (2001); who opined that farming experience shows that farmers will be able to make sound decisions as regards resources allocation and management of their farms.

Access to Agricultural Credit

The estimated coefficient of Credit access (-0.472) was negative and statistically significant at P<0.05 (5%) level of probability. Credit access revealed a positive relationship with technical efficiency. This implies that the farmers who have greater access to credit tend to be more efficient in production; hence they possess more capital for investments in farm assets. This agrees with the study carried out by Onyenweaku et al., 2004 and Belon et al., 2003 who reported similar findings of variables in the inefficiency effects model.

Extension Contact

The estimated coefficient of extension contact (-0.733) was negative and statistically significant at P< 0.05 (5%) level of probability. Extension contact leads to more efficient transmission of information to farmers as well as enhancing the adoption of innovation. Extension contact is germane to build farmers capacity, resilience to agricultural risks and reduction strategies, improve access to innovation and practices that will ensure sustained farm efficiency and food security in the study area. This agrees with the study carried out by Amaza and Maurice, 2005 and Cechura et al., 2014 who reported similar findings of variables in the inefficiency effects model.

Cooperative Membership

The estimated coefficient of Cooperative membership (-0.396) was negative and statistically significant at P< 0.05 (5%) level of probability. Membership of cooperatives is positively and significantly related to technical efficiency. Membership of cooperatives enables the farmers to have access to agricultural information, credit and other inputs as well as enhanced ability to adopt innovations. This suggests that assumed benefits will only accrue to members of co-operative societies who may have pooled their resources together for expansion, efficiency and effective farm management. In a study carried out by Etim et al., (2005) and Cechura et al., (2014) Stated that membership of cooperative societies gives advantages of accessibility to microcredit, input subsidy and agricultural technology. Also, it serves as avenue for the exchange of ideas and agricultural information.

Technical Efficiency Index

The technical efficiency of sampled farmers was less than one (<100%) implying that all the farmers in the study area are producing below maximum efficiency frontier. From the observed range of technical efficiency across the sampled farmers, the best farmer had a technical efficiency index of 0.91 (91%), while the least farmer had a technical efficiency of 0.33 (33%). A wide gap exists between the efficiency of most technically efficient farmer (91%) and that of the least efficient farmer (33%). This type of wide variation in farmer’s specific efficiency levels is a common phenomenon in developing countries (Amaza, 2007). The result, however, indicates that great potential exists for the sweet potato farmers to further increase output through increased efficiency in resource utilization. The mean technical efficiency was 0.62 (62%) implying that on the average, farmers in the study area were able to obtain average of 62% optimal output from a given mix of production inputs. The mean technical efficiency estimated indicates that the realized output could be increased by about 38% by adopting improved management practices and technology. The magnitude of the mean technical efficiency of the farmers is a reflection of the fact that most of the sampled farmers carry out sweet potato production using poor management practices; with inadequate adoption of modern agricultural technology. From this estimation, maximum technical efficiency is not yet achieved suggesting a need for more effort at improving
efficiency of sweet potato farmers in the study area. The distribution of technical efficiency index of the farmers shows that, most (45.7%) of the sweet potato farmers had efficiency index ranging between 0.41-0.50; this further emphasizes the need for the respondents to adopt improved management practices and technology that can boost their level of farm output and efficiency (Cechura, 2014). Thus, opportunity still exists for increasing productivity and income through increased efficiency in resource utilization.

Table 20. Distribution of Respondents based on their Technical Efficiency Index

| Efficiency index | Frequency | Percentage (%) |
|------------------|-----------|----------------|
| 0.31–0.40        | 10        | 10.6           |
| 0.41–0.50        | 43        | 45.7           |
| 0.51–0.60        | 17        | 18.1           |
| 0.61–0.70        | 12        | 12.8           |
| 0.71–0.80        | 7         | 7.4            |
| 0.81–0.90        | 3         | 3.3            |
| 0.91–1.00        | 2         | 2.1            |
| Total            | 94        | 100            |

Minimum 0.33
Maximum 0.91
Mean 0.62
Source: Computed from MLE results 2017

Table 21. Constraints of sweet potato production

| Constraints                                  | F  | %  |
|----------------------------------------------|----|----|
| High cost of labour                          | 65 | 69.1|
| Financial constraints                        | 82 | 87.2|
| Poor storage facility                        | 71 | 75.5|
| Pest and diseases                            | 46 | 48.9|
| Poor access to agricultural technology/inputs| 53 | 56.4|
| Lack of extension contact                    | 34 | 36.2|
| Low patronage due to predominance of similar crop(s)| 37 | 39.4|
| High cost of agricultural technology/inputs  | 89 | 94.7|

F: Frequency, Percentage (%), Source: field survey 2017; * Multiple responses allowed

Contrasts of Sweet Potato Production

The result of Table 21 revealed that the constraints of sweet potato production in the study area include; high cost of agricultural technology/inputs (94.7%); attributable to poor access and non-subsidization of productive resources. Also, most of the farmers wish to wait for government subsidized and qualitative fertilizer and inputs which are grossly inadequate. Financial constraints (87.2%); attributable to poor access to financial institutions and agricultural credit among the respondents; hence their meager savings are not sufficient to cater for their farm activities in sweet potato production. Poor storage facilities (75.5%); attributable to poor access to modern agricultural technology, high cost of labour (69.1%); attributable to non-availability and inadequate supply of farm labour; family labour was predominantly used in the study area resulting in acute shortage of labour. According to the farmers, during active period of production-every household would have been engaged in his family farm work. The demand for labour is normally very high and expensive during the peak period of land clearing, ridging, harvesting, processing and weeding in the study area. Poor access to agricultural technology/inputs (56.4%); attributable to non-availability of agricultural technology/inputs in the study area, according to the respondents they make use of seeds from their previous harvest which is not reliable and can jeopardize improved and sustainable productivity. Pest and diseases (48.9%); attributable to the adoption of poor management practices among the respondents, the farmers also revealed that pest and diseases were responsible for pre-harvest and post-harvest losses in sweet potato production in the study area. Low patronage due to predominance of similar crop(s) (39.4%); attributable to increased demand for alternative root crops in the study area and poor access to market linkages to sell their agricultural produce. Lack of extension contact (36.2%); attributable to poor and inadequate extension services in the study area. All the constraints identified by the farmers significantly affected sweet potato production in the study area. This result is in line with the study carried out by Cechura et.al. (2014); Ogundari and Ojo (2006); and Ben-Belhassen (2002), who opined similar constraints in crop production.

Conclusion and Recommendations

This study analyzed the technical efficiency of sweet potato production in Bokkos LGA of Plateau State, Nigeria. The results revealed that the socioeconomic characteristics of the respondents significantly affected their farm efficiency and level of farm output in sweet potato production in the study area. The variables in the stochastic frontier model significantly affected the technical efficiency and output of sweet potato production in the study area. This implies that efficient resource utilization in sweet potato production can increase the level of farm output in the study area. The efficiency index of the sweet potato farmers in the study area also revealed that they were not producing at optimal capacity. The constraints identified significantly affected sweet potato production in the study area. Based on the findings of this study, the following recommendations are made for policy actions to improve the technical efficiency of sweet potato production in the study area;

- Policies should be formulated that will subsidize agricultural input costs.
- Stakeholders should increase measures to sensitize farmers on appropriate application and utilization of agricultural resources.
- Formulation of policies to improve access to agricultural inputs and technology in sweet potato production.
- Formulation of policies that will ensure adequate market linkages for their agricultural produce.
- Formulation of policies that will ensure adequate farm labour supply.
- Formulation of polices to improve access to farm capital, agricultural credit and extension services for the sweet potato farmers.
- Sensitization of the sweet potato farmers on appropriate methods of fertilizer and agrochemical application and management.
• Formulation of polices that’ll improve the level of farm efficiency in sweet potato production in the study area.
• Sensitization of sweet potato farmers on modern agricultural practices, importance of cooperative formation and efficiency in resource utilization.
• Formulation of policies to encourage the development of indigenous storage and processing technologies in sweet potato production.

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