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Estimation of technical efficiency of black cumin (*Nigella sativa* L.) farming in northwest Ethiopia: a stochastic frontier approach



Abebe Birara Dessie^{1*}, Tadie Mirie Abate¹, Betelhem Tsedalu Adane¹, Tiru Tesfa² and Shegaw Getu²

*Correspondence:
a.birara@yahoo.com

1 Department
of Agricultural Economics,
College of Agriculture
and Environmental Science,
University of Gondar, P.O. BOX
196, Gondar, Ethiopia
Full list of author information
is available at the end of the

Abstract

Ethiopia is one of the east African countries which produce and exports various spices to other countries. Black cumin (Nigella sativa L.) is an important stiff annual flowering plant which mainly grows by producers for its seeds. An increasing demand of black cumin seed and oil in local, national and international market for medicinal, consumption and commercial purpose makes the best alternative crop for small holder farmers in Ethiopia. In spite of its importance, not much has been done to improve its production and productivity in Ethiopia. Therefore, this research was designed to examining efficiency variations and factors influencing technical inefficiency levels of producers on black cumin production in northwest Ethiopia. Primary data were collected using a semi-structured questionnaire administered on 188 black cumin producers selected using systematic random sampling technique. Moreover, various data analysis methods such as descriptive statistics and stochastic frontier model were used for analyzing the data. The empirical result obtained by applying maximum likelihood estimate of stochastic frontier model revealed that seed (p < 0.01) labor (p < 0.05), chemical (p < 0.01) and land (p < 0.05) were significant input variables in determining black cumin production. The mean technical efficiency level of black cumin producer was generally low, about 53.1%. The mean value of actual yield, potential yield and yield gap was 3.131, 5.832 and 2.701 quintals, respectively. Moreover, the result of stochastic frontier model together with the inefficiency parameters revealed that market price of black cumin (p < 0.01) and access of extension service (p < 0.1) were significant variables and positively influenced the efficiency levels of black cumin producers. Whereas age of producers (p < 0.05) and distance to farm plot (p < 0.01) negatively influenced the technical efficiency levels of black cumin producers. Therefore, the study recommends that adoption of latest agricultural technologies; development of institutions, agricultural extension services and infrastructure are advisable to improve the efficiency and commercial value of black cumin production.

Keywords: Black cumin, Efficiency, Stochastic frontier model, Production, Ethiopia

1 Introduction

In developing countries, larger sections of the populations are depends on agriculture sector. In Africa, there is abundance of resources and high potential of attracting investors and producers in various agricultural productions and emerging markets (Fiergbor



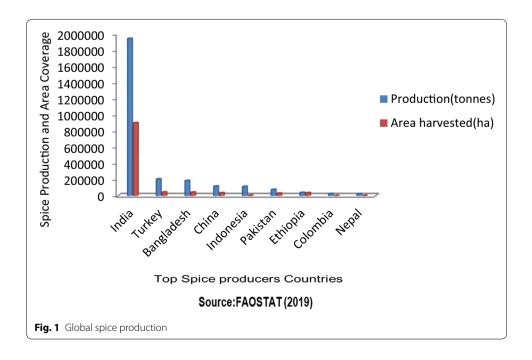
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Dessie et al. Economic Structures (2020) 9:18 Page 2 of 14

2020). However, the sector's production, productivity and efficiency status is low in most African countries including Ethiopia (FAO and WFP 2012). Therefore, adopting of any production technologies are important tool for reducing poverty and production cost (Kassie et al. 2011), smoothing variation and shocks of any country's aggregate output and economy (Ogun 2020) and increase a source of diversified income (Maertens and Barrett 2012). A spice is a dried seed, fruit, root, bark or vegetative agricultural product which produced for consumption and commercial purpose (Dessie et al. 2019a), flavoring of foods, and sometimes as a preservative of food by killing and preventing the growth of some harmful bacteria (Tesfa et al. 2017; Spice Sector Strategy Coordinating Committee 2010). Figure 1 shows that the top nine spice producer countries in the world were India which produces about 1,939,000 tonnes followed by Turkey (199,018 tonnes), Bangladesh (180,993 tonnes), China (113,359 tonnes), Indonesia (110,387 tonnes), Pakistan (73,472 tonnes) and Ethiopia, about 36,754 tonnes (FAOSTAT 2019).

Ethiopia is a country with different and favorable agro ecological zones for production of various spices, vegetables and crops (Dessie et al. 2019a, b). The country grows more than 14 spice varieties particularly black cumin, white cumin, pepper, paprika, turmeric, fenugreek, garlic, korarima, coriander, ginger, cardamom, and basil for consumption and commercial purposes (Etissa 1998; Girma et al. 2008; Tesfa et al. 2017; NABC 2019; Dessie et al. 2019a, b). Similarly, Ethiopia is one of the top spice producer and consumer countries (Vijayalaxmi and Sreepada 2014) and ranking first and seventh in Africa and global spice production, respectively (FAOSTAT 2019). Moreover, Ethiopian societies use spice products for various purposes such as to flavor bread, butter, meat, soups, vegetables, and to medicines and perfumes (International Trade Centre 2010; Tesfa et al. 2017).

Cumin (*Cuminum cyminum*) is an important spice type which is mainly cultivated for the purpose of flavoring vegetables, pickles, soups, sauces, cheese and for pleasant



Dessie et al. Economic Structures (2020) 9:18 Page 3 of 14

aroma (Verma and Kumar 2015; Atta 2003; Takruri and Dameh 1998). About 70% of the world cumin production is produced by India and consumed by China, Indonesia, Singapore, Malaysia, Bangladesh, Nepal and India itself (Spices Board India 2009). Similarly, cumin has two best varieties such as white and black. In 2012, the global cumin production was estimated to be 300,000 tonnes, with India contributing about 210,000 tonnes cumin output (Sonke and Lisanne 2013).

Black cumin (*Nigella sativa* L.) is a flowering plant which belongs to the family Ranunculaceae (buttercup). The scientific term of *Nigella sativa* has other synonyms such as *Nigella damascena*, *Nigella ciliaris*, *Nigella arvensis* and *Nigella hispanica*. *Nigella sativa* is also known by its vernacular names such as black cumin (English), Tikur azimude (Ethiopia), Cheveux de vénus (France), Jinten hitam (Indonesia), Kanauji (Punjabi), Habbet as-suda (Arabic), Jintan hitam (Malaysia), Hak jung chou (Chinese, Cantonese) and Thian dam in Thailand (Alemu 2016; Burits and Bucar 2000). Literatures also revealed that black cumin is native to the Mediterranean region but currently, it is cultivated in different parts of the world including Ethiopia. In recent times, many medicinal values have been accredited to black cumin seeds and its oil.

In Ethiopia, black cumin is one of the most important spice types which are mainly produced to flavor foods, preparation of oil for perfumes and medicinal purpose, source of income, crop diversification, and export purposes (Anshiso and Teshome 2018; Teshome and Anshiso 2019). It is also used for reducing the hotness of pepper powder in the country (Edwards et al. 2003). Therefore, black cumin production can be increased either using modern agricultural technologies or improving the efficiency of inputs with existing technologies. Moreover, Kopp (1981) defined efficiency as the ability to produce the maximum amount of outputs using the existing levels of inputs and technologies.

In the country, the annual production of black cumin was 18,000 metric tons in 2014/2015 crop production season (Ethiopian Investment Agency 2015). The national average productivity of black cumin was 0.79 tonnes/ha (Kifelew et al. 2017). The demand of black cumin seed and its oil has also been increasing both in Ethiopian local and national markets for consumption purpose. It is also the second important cash crop which is exported to international market next to ginger (Teshome and Anshiso 2019).

Even though the production and area coverage of black cumin have been increasing, its productivity is still very low in the country. Yosef (2008) revealed that the main factors influencing the production and productivity of black cumin were lack of improved seed, absence of recommended fertilizer rate; postharvest handling problem; lack of improved agriculture practices and extension system, marketing problems. In Ethiopia, the adoption of modern and intensive agricultural technologies such as the use of row planting, inorganic fertilizer and improved seeds was also very low (Chanyalew et al. 2010). Moreover, the production and productivity of cumin were influenced by various socio-economic, demographic, institutional and environmental factors (Teshome and Anshiso 2019; Nejad and Khorramabad 2011; Lindara et al. 2004).

In Ethiopia, the main black cumin producer regions are Amhara, Oromia, SNNP and Gambiella (Teshome and Anshiso 2019). Though there is the high potential in black cumin production in the country, how much to produce and when to produce, what are the efficiency levels of producers and which factors influence the technical efficiency

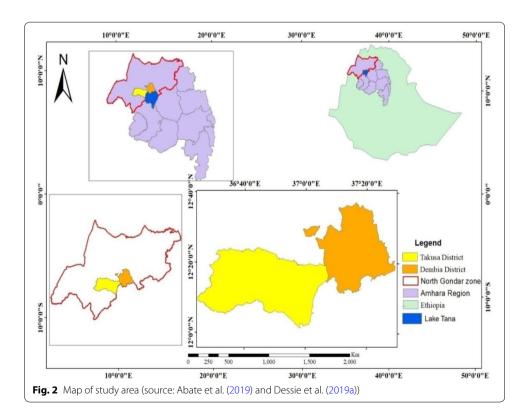
Dessie et al. Economic Structures (2020) 9:18 Page 4 of 14

levels of black cumin producers were hardly studied in the study area. Based on the above statement, the study was intended to empirically answer the following three key research questions: (1) Are the producers technically efficient in black cumin production? (2) What factors influence the technical efficiency levels of black cumin producers? (3) What are the comparison levels of potential and actual black cumin production in the study area? Therefore, studying the efficiency level of producers is imperative to determine the extent to raising black cumin production and productivity using the existing levels of inputs and technology. The findings from this study are believed to be helpful in reducing the information gap on black cumin and contributing to work better on input usage, production and marketing strategies for the benefit of producers, traders, government and ultimate consumers.

2 Research methods

2.1 Description of study area

The study was done in north Gondar zone but currently the zone divided into three zones namely west Gondar, central Gondar and north Gondar zone. The zone is located in the northwest of Ethiopia and 738 km far from the capital city of the country. The capital city of the zone is Gondar city which located at 12° 35′ 60.00″ N latitude and 37° 28′ 0.01″ E longitudes with mean altitude of 2133 m above sea level. In the zone, the main sources of livelihood for households are crop production, vegetable production, animal production, bee keeping and spice production particularly pepper, ginger, white and black cumin. The low land of the zone is dominated by semi-arid natural forest. In the zone, 51% and 49% of population are men and women, respectively (Dessie et al. 2019a,



Dessie et al. Economic Structures (2020) 9:18 Page 5 of 14

b; Abate et al. 2019). The survey was done in 2019 cropping season on two large black cumin producer districts of the zone specifically, Takusa and Dembia districts (Fig. 2).

2.2 Sampling technique and sample size determination

A multi-stage sampling technique was used to select sample producers. In the first-stage, two main black cumin producer districts namely, Takusa and Demebia district, were selected using purposive sampling technique due to its high potential black cumin production. In the second stage, nine main black cumin producer kebeles/rural villages/namely, Mekonta, Chemera, Seravadelo, Guarhie, Tezeba, Darna, Banbaro, Jangua and Gebaba-salge, were also selected using purposive sampling technique due to the best producers' experience in production and marketing of black cumin output. In the third stage, 188 sampled black cumin producers were selected using a systematic simple random sampling technique proportionate to size sampling methodology following (Cochran 1977).

$$n = \frac{Z^2 pq}{e^2},\tag{1}$$

where n is the sample size; Z is the confidence level (Z=1.96); p is the proportion of population containing major interest (p=0.5 and q=1-p) and e is allowable error (e=0.072).

2.3 Methods of data collection

In this study, both primary and secondary data were collected from various sources. The main primary data collection methods were semi-structured questionnaires and key informant's interviews. Moreover, to enrich the investigation secondary data were collected from various sources. The interview schedule which consists of semi-structured questions was prepared in English and translated into local language to collect information on socio-economic, demographic and institutional characteristics of households. Furthermore, the questionnaire was pre-tested using pilot survey, and the necessary amendment was made before the actual survey. Finally, semi-structured interviews were administered on 188 black cumin producers.

2.4 Methods of data analysis and model specification

Various data analysis methods such as descriptive statistics and econometric model were used to analyze the data. Descriptive statistics such as means, frequencies, and percentages were used to examine the socio-economic, institutional and demographic characteristics of sampled producers.

In addition, the most common production functional model such as Cobb–Douglas function was used to measure the physical relationship between production inputs and black cumin output. As compared to transcendental logarithmic (translog) function, the Cobb–Douglas production function is the simplest, less susceptible to multicollinearity problem among explanatory variables and provides coefficients of a plausible sign and magnitude (O'Neill et al. 1999). Moreover, the Cobb–Douglas has been widely used in

Dessie et al. Economic Structures (2020) 9:18 Page 6 of 14

many empirical studies particularly those related to developing countries for farm efficiency analysis (Bravo-Ureta and Pinheiro 1997).

The general form of Cobb-Douglas production function is given in Eq. (2).

$$Y = F(X_i\beta) \exp(V_i - U_i)$$
, where $i = 1, 2, 3 \dots 188$. (2)

Using a linear representation, the empirical production function to be estimated as follows in Eq. (3):

$$\ln Y_i = \beta_0 + \beta_1 \ln \ \text{lab} + \beta_2 \ln \text{land} + \beta_3 \ln \text{seed} + \beta_4 \ln \text{oxen} + \beta_5 \ln \text{chemical} + \beta_6 \ln \text{fert} + \nu_i - u_i,$$
(3)

where Y_i is the production of black cumin yield of ith producer, X_i a vector of inputs used by the ith producer, β a vector of unknown parameters, and v_i a random variable which is assumed to be $N\left(0,\sigma_{v_i}^2\right)$ and independent of the u_i which is nonnegative random variable assumed to account for technical inefficiency in black cumin production.

The farm-specific technical efficiency (TE) is also defined in terms of observed output (Y_i) to the corresponding frontier output (Y_i^*) using the available technology to be estimated in Eq. (4).

$$TE_i = \frac{Y_i}{Y_i^*} = \frac{\text{Actual yield}}{\text{Potential yield}}$$
 (4)

Technical efficiency takes value on interval (0,1), where 1 indicates a fully efficient farm.

In this study, the yield gap (YG_i) of ith producer in black cumin production is the difference between potential yield (PY_i) and actual yield (AY_i) and estimated in Eq. (5) as follows:

$$YG_i = PY_i - AY_i. (5)$$

All parameters of the stochastic frontier and inefficiency effect model were estimated in a single-stage maximum likelihood (ML) method through Frontier 4.1 software following (Coelli 1996). Therefore, a stochastic frontier approach was used to estimate the technical efficiency levels of cumin producers and to differentiate inefficiency from deviations that are caused from factors beyond the control of producers estimated in Eq. (6) as follows:

$$\mu_{i} = \delta_{0} + \delta_{1i} (\mathrm{Age}_{i}) + \delta_{2i} (\mathrm{Educ}_{i}) + \delta_{3i} (\mathrm{TLU}_{i}) + \delta_{4i} (\mathrm{Landzize}_{i}) + \delta_{5i} (\mathrm{Market\ price}_{i}) \\ + \delta_{6i} (\mathrm{Land\ fragementation}_{i}) + \delta_{7i} (\mathrm{Soil\ fertility}_{i}) + \delta_{8i} (\mathrm{extension\ acess}_{i}) + \delta_{9i} (\mathrm{distance\ to\ farmplot}_{i}),$$

$$\tag{6}$$

where μ_i is the technical inefficiency indices; the subscript i indicates the ith household in the sample (i = 1, ..., 188); and $\delta_0, \delta_{1i}, ..., \delta_{9i}$ are parameters to be estimated.

3 Results and discussion

3.1 Demographic and socio-economic characteristics of households

In the study area, about 96.28% and 3.72% sample households were male and female headed, respectively. Similarly, about 57.45% and 42.55% of sample household were illiterate and literate, respectively. Moreover, about 61.17% and 46.81% of producers

Dessie et al. Economic Structures (2020) 9:18 Page 7 of 14

Table 1 Mean and proportion households' characteristics (N = 188)

Continuous variables	Mean	Standard deviation	
Age in years	48.02	0.77	
Family size in AE	6.20	0.16	
Land size allotted to black cumin in ha	0.40	0.20	
Output in quintals	3.13	0.15	
Market price in ETB/Qt	3608.12	53.47	
Distance to development center in km	3.37	0.13	
Distance to the nearest market in km	4.40	0.22	
Dummy variable	Response	Frequency	Percentage
Extension service	Yes	89	47.34
	No	99	52.66
Sex	Male	181	96.28
	Female	7	3.72
Education status	Illiterate	108	57.45
	Literate	80	42.55
Market information	Yes	115	61.17
	No	73	38.83

Table 2 Black cumin yield gap due to technical inefficiency

Variable	Mean	Std. dev.	Min	Max
Actual yield (quintal/0.4 ha)	3.132	2.092	0.800	20.000
TE	0.531	0.212	0.140	0.935
Potential yield (quintal/0.4 ha)	5.832	2.399	2.767	23.625
Yield gap (quintal/0.4 ha)	2.701	1.622	0.519	10.392

were got market information and extension service, respectively (Table 1). The result also revealed that the mean age of households, family size, land size allotted to black cumin, distance to development center, output, market distance, and market price were 48.02 years, 6.2 in AE, 0.4 ha, 3.37 km, 7.825 Qt/ha, 4.4 km, and 3608.12 ETB/Qt, respectively (Table 1). The mean age 48.02 years implies majority of the black cumin producers are still in their active age and thus expected to be productive. This result is in line with the findings of Dessie et al. (2019a, b) and Wassihun et al. (2019) confirmed that the mean age of red pepper and potato producers in northwest Ethiopia was 48.31 and 46.69 years, respectively.

3.2 Black cumin efficiency scores and yield gap due to inefficiency

The result in Table 2 shows the mean technical inefficiency was 46.9% which caused 2.701 quintal yield gap of black cumin which is equivalent to 6.753 Qt per ha. The result also shows the mean value of the actual output and the potential yield of black cumin was 3.132 and 5.832 quintals, respectively. This implies that sample producers in study area were producing on average 2.701 quintal of black cumin yield lower than their potential yield. Similarly, the result indicated that in the short run there is a room to improving efficiency level of producers and yield of black cumin by 2.701

Dessie et al. Economic Structures (2020) 9:18 Page 8 of 14

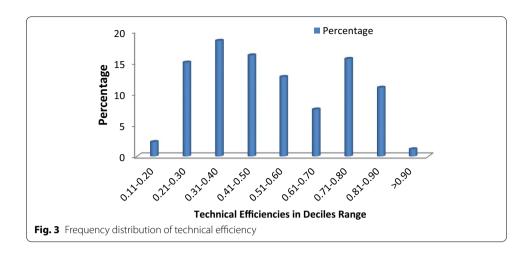


Table 3 Generalized likelihood ratio test of hypotheses for model and parameters

Null hypothesis	Degree of freedom	LR	x² value	Decision
$H_0: \gamma = 0$	1	72.66	3.84**	Rejected
$H_0: \beta_7 = \cdots = \beta_{27} = 0$	21	30.53	32.67**	Not rejected
$H_0: \delta_0 = \cdots = \delta_9$	9	72.66	16.92**	Rejected

^{**}Statistically significant at 5%

quintal using the existing levels of input and technology (Table 2). This finding is in line with Abate et al. (2019) who found that the mean technical inefficiency and yield gap of red pepper in northwest Ethiopia were 21.2% and 191.452 kg, respectively. Similarly, Wassihun et al. (2019) confirmed that the mean technical inefficiency of producers in potato production was 54% in northwest Ethiopia.

The result in Fig. 3 shows the frequency distribution of technical efficiency levels of black cumin producers. In the study area, the technical efficiency levels of black cumin producers vary from one to another. The efficiency levels of growers range from 0.140 to 0.935. The majority of cumin producers (18.49%) have a lower technical efficiency score (between 0.31 and 0.40). The result in Fig. 3 also revealed that the distribution of the TE scores is skewed to the left implying less technically efficient. More than 16.18% of sampled producers have also a TE score between 0.41 and 0.50. The frequency distribution of efficiency indexes indicates there is a high technical efficiency variation among producers. Therefore, the presence of technical inefficiencies could be eliminated by implementing the practice of technically efficient black cumin producers in the study area. This finding is confirmed with Lindara et al. (2004) endorsed that the mean technical inefficiency of large spice producers and small spice producers in Sir Lanka was 20.57% and 10.07%, respectively.

3.3 Estimation of production function and hypothesis tests

The stochastic production frontier was applied using the maximum likelihood estimation procedure using Frontier 4.1 computer program. Prior to model estimation, all the hypotheses (assumption of stochastic frontier) were tested using generalized likelihood

Dessie et al. Economic Structures (2020) 9:18 Page 9 of 14

ratio (LR). The LR test was made to decide whether the traditional average production function (OLS) best fits the data set as compared to the stochastic frontier model (SFM). The test was carried out by estimating the stochastic frontier production function and conducting a likelihood ratio test assuming the null hypothesis of no technical inefficiency ($H_0: \gamma=0$). The likelihood ratio test statistics was 72.66. This value exceeds the critical $x^2(5\%,1)$ value of 3.84 at 5% level of significance (Table 3). The result shows the null hypothesis was rejected and implying there is statistically significant inefficiency in the data. This also implies that stochastic frontier production function was an adequate representation of the data.

Likewise, the LR statistical test was estimated for the selection of the appropriate functional form (Cobb–Douglas versus Translog production function). The calculated likelihood ratio (LR) is equal to 30.53 and the critical value of x^2 at 21 degrees of freedom and 5% significance level is 32.67 (Table 3). Thus, the null hypothesis that all coefficients of the interaction terms in Translog specification are equal to zero was not rejected. This implies that the Cobb–Douglas functional form sufficiently represents the data. Hence, the Cobb–Douglas functional form was used to estimate the technical efficiency levels of the black cumin producers in the study area.

Moreover, LR statistical test was estimated to test the third assumption. Farm-level technical inefficiencies are not affected by the socio-economic variables included in the inefficiency model ($H_0: \delta_0 = \delta_1 = \cdots = \delta_9 = 0$). The calculated LR value of 55.59 was greater than the critical value of 16.92 at 9 degrees of freedom. The result shows that the null hypothesis (H_0) that explanatory variables are simultaneously equal to zero was rejected at 5% significance level. This implies that there is a room to improving black cumin output by identifying those institutional, socio-economic and farm-specific factors causing this variation.

Furthermore, the degree of multicollinearity among the explanatory variables has been tested using variance inflation factor (VIF) test for continuous variables and contingency coefficient (CC) test for dummy variables. Therefore, the post estimation

Table 4 OLS and ML estimate of production frontier for black cumin producers

Variables	Parameters	Ordinary least square estimates		Maximum likelihood estimates	
		Coefficient	Std. err	Coefficient	Std. err
Constant	$oldsymbol{eta}_0$	0.547*	0.324	0.683**	0.345
Lnseed	$oldsymbol{eta}_1$	0.190**	0.080	0.332***	0.097
LnNPS	$oldsymbol{eta}_2$	0.032	0.026	-0.043	0.031
Lnland	$oldsymbol{eta}_3$	0.240**	0.097	0.227**	0.103
LnChemical	eta_4	0.251***	0.047	0.185***	0.049
LnOxen	$oldsymbol{eta}_5$	0.049	0.114	0.052	0.112
LnLabor	$oldsymbol{eta}_6$	0.159*	0.085	0.202**	0.080
Sigma-squared (σ^2)				0.211***	0.036
Gamma (γ)				0.849***	0.076
Return to scale				0.946	
Log likelihood function		- 127.582		- 91.247	
Total sample size		188		188	

^{***, **} and * are statistically significant at 1%, 5% and 10%, respectively

Dessie et al. Economic Structures (2020) 9:18 Page 10 of 14

result shows the value of variance inflation factor and the contingency coefficient were less than 10 and 0.75, respectively. This implies that the data have not multicollinearity problems.

One of the attractive features of the Cobb–Douglas production functional form is the direct interpretation of its parametric coefficients with respect to the level of input used. This feature allows researchers to evaluate the potential effects of changes in the amount of production input used on the amount of output to be produced.

The result of stochastic frontier model in Table 4 revealed that four input variables such as seed, land, chemical (herbicide and pesticides), and labor positively and significantly influenced the amount of black cumin production. This implies that as the amount of seed, size of land, amount of chemical, and number of labor increased by 1%, the amount of black cumin production also increased by 0.332%, 0.227%, 0.185%, and 0.202%, respectively (Table 4). This study is in line with recent studies of Abate et al. (2019), Wassihun et al. (2019), Tefaye and Beshir (2014)and Lindara et al. (2004); they revealed that the amount of input usage has a direct implication on crop and spice production.

The value of sigma-squared (σ^2) for frontier of black cumin output was 0.211 and statistical significant at 1% level. The significant value indicates the goodness of fit of the specified assumption of the composite error terms distribution (Ahmed 2014; Kifle et al. 2017; Abate et al. 2019). The estimated value of gamma was also 0.849. This implies 84.9% of total variation in black cumin output was due to technical inefficiency of producers.

Moreover, the return to scale of this study was estimated to be 0.946% implying decreasing returns to scale. This implies that a 1% increase in all inputs proportionally would increase the total production of black cumin by less than 1%. This result is in line with the findings of Abate et al. (2019) who found that the estimated return to scale was 0.983% in red pepper technical efficiency farming systems of northwest Ethiopia. Alemu et al. (2009) also endorsed that the estimated return to scale was 0.962%, implying decreasing return to scale across technical efficiency farming systems in Ethiopia.

3.4 Determinants of technical inefficiency of black cumin producers

Age of producer was statistically significant (p < 0.05) and positively influences the technical inefficiency of black cumin production at 5% of level of significance (Table 5). The

Table 5 Maximum likelihood estimates of technical inefficiency determinants

Variables	Parameters	Coefficient	Std. err	<i>T</i> -ratio
Constant	δ_0	11.935***	2.497	4.779
Age	δ_1	0.010**	0.004	2.299
Education status	δ_2	- 0.134	0.108	-1.242
Tropical livestock unit	δ_3	0.002	0.016	0.137
Land size allotted to cumin	δ_4	0.073	0.276	0.265
Market price	δ_5	- 1.440***	0.312	-4.611
Land fragmentation	δ_6	-0.024	0.133	-0.180
Soil fertility	δ_7	- 0.121	0.175	- 0.689
Extension service	δ_8	− 0.274*	0.145	- 1.887
Distance to farm plot	δ_9	0.074***	0.026	2.846
Log likelihood function		- 91.247		

^{***, **} and * are statistically significant at 1%, 5% and 10%, respectively

Dessie et al. Economic Structures (2020) 9:18 Page 11 of 14

significantly, positive relationship between age and technical inefficiency of black cumin production implies that older producers are more technical inefficient than younger producers. This can be explained by the fact that as a producer become more aged, it becomes difficult for proper utilization of inputs and adoption of latest technologies on their farms. Likewise, younger producers will be more energetic, active and can make well-thought decisions in farming system and as a result, their efficiency level in black cumin production increases. This result is consistent with the recent findings (Kinde Teshome 2005; Dolisca and Jolly 2008; Amodu et al. 2011; Hailemaraim 2015; Haile et al. 2018; Dessie et al. 2019a, b; Begum et al. 2019).

Market price of black cumin was statistically significant (p < 0.01) and negatively influences the technical inefficiency of black cumin production at 1% level of significance. The negative relationship between market price and technical inefficiency implies that the technical inefficiency level of producers decreased as the commercial value of black cumin increases.

Extension service was statistically significant (p<0.1) and negatively influences the technical inefficiency of black cumin production at 10% level of significance (Table 5). The negative relationship between extension service and technical inefficiency implies farmers who got extension advice and training was less technically inefficient than those who did not accessed. It also implies that extension advice and training improve both farmers' ability of choosing appropriate and least cost combination of production inputs and ability to producing the optimum amount of yield using such inputs. Moreover, agricultural extension services are supposed to improve the efficiency level by closing the gap between knowledge of farmers and the existing technological and technical knowledge. This finding is consistent with Gelaw and Bezabih (2004), Begum et al. (2019) and Abate et al. (2019); they endorsed that a well-managed extension service is expected to improve farmers' efficiency level.

Distance to farm plot was statistically significant (p<0.01) and positively influences the technical inefficiency of black cumin production at 1% level of significance (Table 5). The significant and positive relationship between farm plot distance and technical inefficiency implies that as the distance to farm plot increases, the technical inefficiency level of black cumin producers also increases. This could be attributed to the fact that the farther the farm plot from producer's residence, the greater would be the cost of transport and poorly managed their farm activities in turn exposed for technical inefficiency.

3.5 Conclusion and recommendations

Black cumin is an important spice crop which is mainly produced by Ethiopian small holder farmers for medicinal value, commercial purpose, and flavorings of food items. However, information on production, productivity and technical efficiency levels of black cumin is limited in the country. Therefore, the main aim of this study was to examine the efficiency variations and factors influencing technical efficiency levels of producers on black cumin production in northwest Ethiopia. Combinations of data analysis methods such as descriptive and econometric analysis were used to analyze the data. The finding revealed that the mean technical efficiency level of black cumin producers was low. It also shows there is large yield gap between actual and potential yield of black cumin in the study area. The study implies that there is an opportunity of increasing the

Dessie et al. Economic Structures (2020) 9:18 Page 12 of 14

levels of black cumin production and productivity by improving the technical efficiency levels of producers using the existing levels of inputs. The return to scale in the study area was decreasing, implying as a black cumin producer uses one unit of input the total black cumin production proportionally increases by less than one unit. The maximum likelihood estimate of stochastic frontier model revealed that important production input variables such as seed, land, chemical (herbicide and pesticides), and labor positively and significantly influenced the black cumin production. The positive coefficient of parameters indicates that as the producers increased the input usage by 1%, the levels of black cumin production also increase by less than 1%. Moreover, the result of stochastic frontier model together with the inefficiency parameters shows market price of black cumin output and extension service were decreased inefficiency whereas age of producers and distance to farm plot were increased inefficiency levels. Based on the findings, the inefficiency levels in black cumin farming can be reduced significantly by strengthening extension services, local and national marketing institutions and providing intensive training to black cumin producers. Generally, for holistic black cumin production and marketing, there should be strong linkage and integration among producers, university centers, agricultural research centers, development agents and policy makers.

Abbreviations

AY: Actual yield; PY: Potential yield; YG: Yield gap; CSA: Central Statistical Agency; e: Error; ETB: Ethiopian Birr; FAO: Food and Agriculture Organization; Ho: Null hypothesis; Ha: Alternative hypothesis; LR: Log likelihood ratio; ML: Maximum likelihood; NABC: Netherlands-Africa Business Center; n: Sample size; OLS: Ordinary least square; SFM: Stochastic Frontier model; SNNPS: Southern Nation and Nationality's People; Qt: Quintal; TE: Technical efficiency; Z: Confidence level; VIF: Variance inflation factor; CC: Contingency coefficient; WFP: World Food Program.

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Authors' contributions

All authors read and approved the final manuscript.

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Availability of data and materials

The author wants to declare that they can submit the data at any time based on publisher's request. The datasets used and/or analyzed during the current study will be available from the author on reasonable request.

Ethics approval and consent to participate

Ethical clearance letters were collected from University of Gondar research and community service directorate and North Gondar Zone administrative office to care for both the study participants and the researchers. During survey, official letters were written for each District and kebele/villages/informed verbal consent was obtained from each client, and confidentiality was maintained by giving codes for each respondent rather than recording their name. Study participants were informed that clients have a full right to discontinue or refuse to participate in the study. Hence, all participants throughout the research, including survey households, enumerators, the supervisors and key informants, were fully informed of the objectives of the study. They were approached friendly in free moods until the end of this research.

Consent for publication

Not applicable.

Competing interests

The authors declare that they have no competing interests.

Author details

¹ Department of Agricultural Economics, College of Agriculture and Environmental Science, University of Gondar, P.O. BOX 196, Gondar, Ethiopia. ² Department of Horticulture, College of Agriculture and Environmental Science, University of Gondar, Gondar, Ethiopia.

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Dessie et al. Economic Structures (2020) 9:18 Page 13 of 14

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