

Energy Efficiency Analysis of White Button Mushroom Producers in Alburz Province of Iran: A Data Envelopment Analysis Approach

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ABSTRACT

The aim of this study was to determine energy consumption pattern and specifically to measure and benchmark the efficiency for white button mushroom production in Alburz province of Iran. The data used in this study were collected by interviewing mushroom producers in the region. In the surveyed farms, average yield and total energy consumption were calculated as around 208.46 kg·ton⁻¹ compost and 133.25 MJ·ton⁻¹, respectively. The results revealed that fossil fuel (40.43%), compost (30.45%) and electricity (27.42%) consumed the bulk of energy. The results of DEA approach also showed that 12 and 14 farmers had efficiency score of unity. Electricity and fossil fuel were found to be used in excess in target mushroom production farms. Moreover, we came to the conclusion that the total energy consumption can be reduced to 120.15 MJ·ton⁻¹ for mushroom production in which diesel fuel energy (50.89 MJ·ton⁻¹), FYM (37.32 MJ·ton⁻¹) and electricity (30.34 MJ·ton⁻¹) energies were considerably significant.

Keywords: Energy; Technical Efficiency; Data Envelopment Analysis; White Button Mushroom Production

1. Introduction

Button mushroom (*Agaricus bisporus*) is regarded as a high protein, low-calorie food with medicinal properties [1]. It is the most widely cultivated, harvested, and distributed mushroom in the world. Mushrooms contain substances of high nutritional value such as minerals and a variety of vitamins. White button mushroom is the most prevalent mushroom in the world [2] and needs to be fresh or dried for various uses in the food industry. Mushrooms also have very low energy levels; five medium-sized button mushrooms added together only have twenty calories (80 kilojoules).

Agricultural sector between other parts has a significant place in producing food requirements of the growing population of the world. On the other hand, agriculture itself uses a large amount of energy in form of inputs in order to produce larger amounts of energy output. Based on these facts, a harmony between resource scarcity and food safety is needed in practice. Namely, sustainability should be introduced to today's agricultural systems. A sustainable agriculture, like all other sustainable devel-

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opment, must meet the needs of the present without diminishing opportunities for the future.

Energy use in agriculture has developed in response to increasing populations, a limited supply of arable land, and a desire for an increasing standard of living. As the population continues to rise, the agricultural sector may face the risk of breaking down in the near future from a lack of energy. However, reducing the dependence on fossil fuels will gain in importance as efforts continue to reduce GHG emissions and as fuel supplies eventually do dwindle. Fortunately, the development of alternative farming methods and advances in biotechnology and mechanization of production systems will likely achieve this end. An example of fossil fuel use in agriculture is to manufacture inputs such as fertilizers, pesticides, and herbicides and also to power industrial farm machinery.

Apart from the above mentioned factors in regard of energy importance, increasing input costs consisting fuel cost, irrigation cost, application of chemicals and transportation cost has caused high a raise in energy costs.

Although considerable data exist in the literature regarding the energy consumption for energy consumption in button mushroom production system such as kiwi [3],

apple [4,5], grapes [6], sunflower [7,8], oilseed [9], greenhouse vegetable production [10,11], little information is available on the analysis of energy consumption in white button mushroom production.

Efficiency was defined by Sherman [12] as the ability to produce the outputs using a required minimum resource level. In production, efficiency is defined as the ratio of weighted sum of outputs to inputs or as the actual output to the optimal output ratio. The weights for inputs and outputs are estimated to the best advantage for each unit so as to maximize relative efficiency. In order to measure the optimal input or output amounts, it is necessary to first specify the production frontier [13].

Iranian agricultural sector needs taking a serious look at the ways of energy expenditure and improving efficiency on farms to reduce their ongoing costs and accordingly improve their bottom line. A wrong belief among farmers causing the inefficient use of energy sources is the excess use of resources to get higher productivity, particularly when they are priced low, free or accessible in plenty [14].

One of the proposed and well-established ways to evaluate the relative technical efficiency (TE) of entities by some mathematical programming models is Data Envelopment Analysis (DEA) [15]. By applying this approach firstly the efficient and inefficient decision making units (DMUs) would be detected and then the optimized amount of inputs use can be decided [16].

In recent years, DEA has become renowned in agricultural researches. Recently, Sefeedpari [16] used DEA to assess the technical efficiency in industrial dairy farms of Tehran province in Iran. In this study the efficiency of farmers based on the constant and variable returns to scale models were found to be 0.88 and 0.93, respectively. It was also concluded that DEA was a useful tool to improve the productive efficiency of farms. In another study [17], DEA was applied to investigate the efficiency of individual farmers and identify the efficient units in citrus production in Spain. Mousavi-Avval et al. [18] optimized energy use and energy costs for apple production using data envelopment analysis. They reported that 54% of farmers are technically efficient. The technical efficiency score was calculated as 0.78. Kiwi production was investigated by Mohammadi et al. [19] from the view of energy efficiency. The technical, pure technical and scale efficiencies of farmers were calculated as 0.942, 0.993 and 0.948, respectively. In spite of the careful literature review on DEA application in agricultural production systems, no study was found with this criterion in edible mushroom production.

Considering the importance of efficient use of energy in Iran, the main objective of the present study was to investigate the energy efficient farms during operations of white button mushroom production in Alburz province of Iran. To achieve this, as the first step, energy use pattern for button mushroom was specified. Also, it ranks efficient and inefficient farmers and sketches the optimum footprints of input use and its potential to energy saving.

2. Material and Methods

2.1. Case Study Region Selection and Data Collection

This study was conducted in Alburz province of Iran within 35°31' and 36°32' north latitude and 50°18' and 51°18' east longtitude with total area of 17,953 km² (1.09% of total country area) [20]. This province is situated 1300 meter above sea levels. Karaj is the central city of this province. Constitution of the farms of agricultural products in this city has caused the upgradability of this sector in recent years. The adjacency to markets of Tehran province and its geographical location can be enumerated as the main reasons of high demand for agricultural and livestock products in Alburz province [21].

Required data used in this study were obtained from all button mushroom producers in Alburz province, Iran. The survey to collect quantitative information was conducted on energy inputs used for the production of white button mushroom in the production period of October 2011-December 2011. In fact, this period is a representative of a production period of button mushroom which takes about two months and the collected data were then converted and so they can be attributed to the average input use in a whole year. To achieve this, a comprehensive and definite questionnaire was provided and completed by farmers. In order to use the needed software programs (DEA Solver), we were supposed to collect sufficient data in accordance to number of input and output parameters (at least 20 DMUs). Hence, questionnaires were sent to the whole white button mushroom production units (26 units).

2.2. Energy Balance Analysis Method

Data collection included the quantity of various inputs use in the form of chemicals, chemical fertilizers, farmyard manure (FYM), diesel fuel, natural gas, electricity, water for irrigation, human labor or machinery and equipment. A standard procedure was used to convert each agricultural input and output into energy equivalents. The energy equivalent may thus be defined as the energy input taking into account all forms of energy in agricultural production. The energy equivalents were computed for all inputs and outputs using the conversion factors (obtained from previous studies) indicated in **Table 1**. Accordingly, the energy equivalents were calculated by multiplying the quantity of the inputs used and output

Table 1. Energy equivalent coefficients of inputs and output in white button mushroom production in Alburz, Iran.

Item	Unit	Energy equivalent (MJ·unit ⁻¹)	References
Inputs			[22]
1. Human labour	h	2.2	
2. Machinery	kg		[23]
Tractor		93.61	[23]
Other machinery		62.7	[24]
3. Diesel fuel	L	56.3	[25]
4. Natural gas	m^3	49.5	
5. Chemicals	kg		[26]
Herbicides		238	[26]
Insecticides		101.2	[26]
Fungicides		216	
6. Fertilizers	kg		[26]
Nitrogen		66.14	[26]
Farm yard manure		0.3	[26]
Compost (kg)	kg	8.034	Calculated
7. Water for irrigation	m^3	1.02	[26]
8. Electricity	kWh	12	[25]
9. Straw	kg	12.5	[24]
Outputs			
White button mushroom	kg	27	[27]

harvested per hectare with their conversion factors.

It should be noted here that energy equivalent of human labor varies considerably, depending on the approach chosen; it must be adapted to the actual living conditions in the target region [28]. In this study the energy coefficient of 1.96 MJ·h⁻¹ was applied. It means only the muscle power used in different field operations of crop production. Also, in order to make an analysis of the embodied energy in the farm machinery, it was assumed that the embodied energy of tractors and agricultural machinery be depreciated during their economical life time [29]; Also, the embodied energy in machinery was calculated by multiplying the depreciated weights of machinery (kg·ha⁻¹) with their energy equivalents (MJ·kg⁻¹) using the following Equation (1) [30]:

$$ME = \frac{G \times M_p \times t}{T} \tag{1}$$

where ME is the machinery energy per unit area (MJ·ha⁻¹); G is the machine mass (kg); M_p the production energy of machine (MJ·kg⁻¹), t is the time that machine used per unit area (h·ha⁻¹) and T is the economic life time of machine (h).

For compost energy, we did not find any energy equivalent coefficient indicating energy content coefficient of compost fertilizer as a significant input in white button mushroom production. For this reason, we in study were supposed to measure this coefficient. For this purpose, the compost production process was thoroughly investigated and its energy use was determined. The energy inputs to compost production were human labour, diesel fuel, electricity, stationary equipment, wheat straw, animal manure, urea fertilizer and water. Here, also, each input use amount has been multiplied by their energy equivalents and finally the energy consumption for each kilogram of compost was found.

2.3. Data Envelopment Analysis Technique

In this study, in order to evaluate the technical, pure technical and scale efficiencies of individual farmers, a nonparametric method of DEA was employed. It is believed that farmers use similar inputs and produce the same product (white button mushroom) and operate in a relatively homogeneous region (e.g., topography, soil type, climatic conditions, etc.). In the above methodology, the energy consumed from different energy sources including: human labor, machinery, diesel fuel, fertilizer, chemicals, water for irrigation, electricity and natural gas were defined as input variables; while, the white button mushroom yield was the single output parameter; also each farmer called a DMU. Data analysis was done by DEA Solver software.

In DEA, two models would be applied in this study including CCR and BCC models. Charnes *et al.* [31] developed the CCR DEA model by and assumes constant returns to scale. On the other hand, the BCC model developed by Banker *et al.* [32] and assumes variable returns to scale conditions. Also applying each model would lead to a specific efficiency value. Efficiency score obtained from CCR model is called technical efficiency and the efficiency value calculated by BCC model is pure technical efficiency. Scale efficiency gives quantitative information of scale characteristics; it is the potential productivity gain from achieving optimal size of a DMU. In Equation (2), scale efficiency is calculated by the relation between technical and pure technical efficiencies as below [33]:

Pure Technical Efficiency =
$$\frac{\text{Technical Efficiency}}{\text{Pure Technical Efficiency}}$$
(2)

In DEA, there are two different analyses for convert-

ing an inefficient unit to efficient one. An inefficient DMU can be made efficient either by minimizing the input levels while maintaining the same level of outputs (input oriented), or, symmetrically, by increasing the output levels while holding the inputs constant (output oriented). In fact, the input-oriented is commonly utilized in DEA applications because efficiency profitability depends on the efficiency of operations. Besides that, a farmer is able to take the control of inputs use more easily than output level [15,34].

Let the DMU_j to be evaluated on any trial be designated as DMU_o ($o = 1, 2, \dots, n$). To measure the relative efficiency of a DMU based on a series of n DMU_s, the model is structured as a fractional programming problem as follows [35]:

Maximize
$$\theta = \frac{\sum_{r=1}^{s} u_r y_{r0}}{\sum_{i=1}^{m} v_i x_{i0}}$$

s.t. Model (1)
 $\frac{\sum_{r=1}^{s} u_r y_{rj}}{\sum_{i=1}^{m} v_i x_{ij}} \le 1, \ j = 1, 2, \dots, n$
 $u_r \ge 0, \ v_i \ge 0$

Using a linear programming (LP) problem, Model (1) can be equivalently written as follows [35]:

Maximize
$$\theta = \sum_{r=1}^{n} u_r y_{ri}$$

st.

$$\sum_{s=1}^{m} v_s x_{si} = 1, i = 1, 2, \dots, k$$

$$\sum_{i=1}^{n} u_r y_{ri} - \sum_{s=1}^{m} v_s x_{si} = 0$$

$$u_r \ge 0, \ r = 1, 2, \dots, n$$

$$v_s \ge 0, \ s = 1, 2, \dots, m$$

where θ is the technical efficiency. Dual linear programming (DLP) problem is simpler to solve than Model (2) due to fewer constraints. Mathematically, the DLP can be written in vector matrix notation [35]:

Maximize
$$\theta$$
s.t.

 $Y\lambda \ge y_0$ Model (3)

 $X\lambda - \theta x_0 \le 0$

where y_o is the $s \times 1$ vector of the value of original outputs produced and x_o is the $m \times 1$ vector of the value of original inputs used by the oth DMU. Y is the $s \times n$ matrix of outputs and X is the $m \times n$ matrix of inputs of all n units included in the sample. λ is a $n \times 1$ vector of weights and θ is a scalar with boundaries of one and zero

which determines the technical efficiency score of each DMU. Model (3) is known as the input-oriented CCR model. It assumes constant returns to scale (CRS), inferring that an increase in inputs would lead in a proportionate increase in outputs.

The results of DEA models divide the DMUs into two sets of efficient and inefficient units; the inefficient units can be ranked according to their efficiency scores. Since DEA lacks the capacity to discriminate among efficient units; a number of methods are in use to enhance the discriminating capacity of DEA [36]. In this study, the benchmarking method was applied to overcome this problem. In this method, an efficient unit which is chosen as a useful target for many inefficient DMUs, and so appears frequently in the referent set, is highly ranked.

For the purpose of estimating the optimized energy input level of each input for inefficient units, λ coefficient was applied. The optimal input use calculation can be defined as follows (Equation (3)):

$$\theta^* x_0 - s^- = X_n \lambda_n \tag{3}$$

where θ^* is technical efficiency score of inefficient units, x_0 energy input of *n*th referent DMU and s^- is input slack of inefficient units compared to efficient DMUs [37].

In this study for data analysis, the Microsoft Excel spreadsheet and the DEA Solver programs were employed.

3. Results and Discussions

3.1. Energy Balance Analysis for White Button Mushroom Production

Prior to energy use pattern analysis of white button mushroom production, energy use amount in compost production (among the most important inputs of white button mushroom production) was estimated. **Table 2** shows the inputs use and their energy use equivalent values.

As the result of energy use pattern analysis, the amounts of energy inputs use, mushroom yield and their energy equivalents were determined. These results are presented in **Table 3**.

As it is obvious from **Table 3**, total energy consumption in white button mushroom production is about 133 MJ/kg of product. Based on the results, fossil fuels, compost fertilizer and electricity are mostly contributed inputs to energy consumption in which fossil fuels (diesel fuel and natural gas) had the first rank (40.43%). Fossil fuel was normally used for heating. It should be noted here that the fossil fuel energy input to compost production process has been focused in our formerly calculations. Moreover, energy input of mushroom seed was not considerable for energy analysis. As a matter of fact, ten kilograms of seed is normally required for one ton of compost. Due to the short planting period of mushroom

Table 2. Mean input use and energy equivalent values in compost production process in Alburz, Iran.

Inputs	Mean input use (unit·ton ⁻¹)	Energy equivalent (MJ·ton ⁻¹)	Share (%)	
Human labour (h)	10.6	23.32	0.29	
Diesel fuel (L)	16.6	934.75	11.64	
Stationery equipment (kg)	10.41	72.78	0.91	
Electricity (kWh)	833.3	2999.88	37.35	
Straw and stubble (kg)	310	3875	48.25	
Poultry manure (kg)	320	96	1.19	
Urea fertilizer	0.003	0.22	0.002	
Water (L)	5	29.4	0.37	
Total input energy (MJ)		8034.44		

Table 3. Mean input use and energy equivalent values in white button mushroom production process in Alburz, Iran.

Items	Mean input use (unit·ton ⁻¹)	Energy equivalent (MJ·ton ⁻¹)	Share (%)	
Inputs				
Compost (kg)	4.86	40.57	30.45	
Fuel		53.87	40.43	
Diesel fuel (L)	0.95	45.5		
Natural gas (m³)	1.1	8.37		
Chemicals	0.0085	1.05	0.78	
Human labour (h)	0.21	0.47	0.47	
Electricity (kWh)	3.05	36.55	27.42	
Equipment (kg)	0.049	0.45	0.37	
Water for irrigation (L)	49.23	0.29	0.22	
Total input energy		133.25		
Output				
White button mushroom (kg)	1	27		

and the need for keeping the temperature at 25 °C, too, there is a high need to fossil fuel use in mushroom production farms for heating purposes. Moreover, since mushrooms do not require oxygen in their first month, air ventilation is not essential unless carbon dioxide content rise from its usual and standard rate. Hence, isolation of planting places would eventually lead in energy saving. Compost was reported as the second high contributing

input to energy consumption amount in this study. The amount of compost use was 4.86 kg per ton of yield (white button mushroom) with energy input of 40.57 MJ·ton⁻¹ (30.45%).

In the study conducted by Ghojebeig [38] in greenhouse products of Tehran province, total energy input use was calculated to be 1.57 MJ for each kilogram of cucumber. They also reported fuel (for heating the greenhouses) and animal manure as the high energy consuming inputs in cucumber production. Moreover, Banaeian et al. [39] investigated energy use pattern in 25 greenhouses of strawberry production in Tehran province, Iran. In our study, results indicated that a total energy of 121891.33 MJ·ha⁻¹ (about 2 MJ·ton⁻¹) was consumed in greenhouse units. Diesel fuel (78%), chemical fertilizers (10%) and electricity (4.5%) were the highest energy consuming inputs. On the contrary, water for irrigation contribution was relatively low (0.47%). This is mainly due to lower water need of mushroom in contrast with other products.

3.2. Data Envelopment Analysis Findings

3.2.1. Technical Efficiency Analysis

Data obtained from questionnaire approach performed in white button mushroom farms were analyzed for the purpose of measuring farmers' technical efficiency in light of energy use by applying data envelopment analysis approach. In the following sections, the efficient and inefficient farmers, their efficiency scores and the required changes for the better resources use management strategies are given. It should also be noted here that both explained DEA models (BCC and CCR models) have been utilized in the present study.

Figure 1 illustrates technical efficiency scores of applied data using CCR and BCC models. The results revealed that, from the total of 26 farmers considered for the analysis, 14 farmers (53.8%) had the pure technical efficiency score of 1. Moreover, from the technically efficient farmers 12 farmers (27.66%) had the technical efficiency score of 1; showing that they were globally efficient and were operating at the most productive scale size of production [40]. In addition, 9 and 7 farmers had their technical efficiency and pure technical efficiency score in the 0.9-1 range, respectively. Both DEA models, reported 5 farmers in the efficiency range of 0.8 - 0.9. It can be derived from Figure 1 that the BCC model reduces the number of feasible units compared to the CCR model and yields a comparatively higher number of efficient DMUs.

Table 4 specifies the mean value of the three efficiency types (technical efficiency (TE), pure technical efficiency (PTE) and scale efficiency (SE)) for inefficient farmers. The results showed that these average values

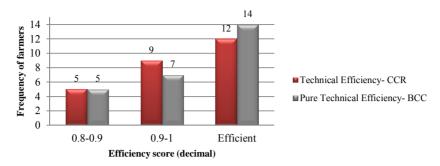


Figure 1. Frequency distribution of farmers from CCR and BCC models in white button mushroom production.

Table 4. Mean value of efficiency types for white button mushroom farmers.

Efficiency type	Mean	SD^*
Technical efficiency	0.955	0.0571
Pure technical efficiency	0.956	0.0573
Scale efficiency	0.999	0.0017

^{*}SD: Standard Deviation.

were 0.955, 0.956 and 0.999 respectively. These inefficient DMUs were in the efficiency range of 0.8 - 0.99. PTE had the highest variation indicating the farmers were not fully aware of the right optimal quantity and production techniques or did not applied them at the proper time.

Omid *et al.* [37] applied the nonparametric approach of DEA to determine the technical, pure technical and scale efficiencies of farmers in some selected greenhouses production in Iran. They considered 18 greenhouse units and 6 of them had the efficiency score of unity. They reported that the TE of the inefficient DMUs, on average, was calculated as 91.5%. In another study [19], the technical, pure technical and scale efficiency of kiwi fruit farmers were calculated as 0.942, 0.993 and 0.948, respectively.

3.2.2. Prioritization of Efficient Units

In this study for ranking the efficient farmers, the number of times they appear in a referent set based on CCR model was counted [36]. The efficient DMUs can be selected by inefficient DMUs as best practice DMUs, making them a composite DMU instead of using a single DMU as a benchmark. The results revealed that farmers No. 13, 9 and 22 with the average cross efficiency scores of 1 had the highest appearance times in referent set (**Table 5**). Therefore, after this step, these farms can be used as terms of benchmarking and establishing the best practice management.

3.2.3. Input Slacks Analysis

We know that when the pure TE score of a producer is

less than one, at present, he is using more energy than required from the different sources. Therefore, it is desired to suggest realistic levels of energy to be used from each source for every inefficient grower in order to avoid energy wastes without reducing the yield level (input-oriented). This can be done by using the value of slacks. Table 6 demonstrates the results of applying BCC model (input-oriented) for estimating the overuses and the slacks for the inputs amount. The slack values indicate that apart from reducing inputs by the amount of $(1-\theta)$, the inefficient DMUs have to reduce their inputs by the amounts indicated by the respective slacks in order to become allocatively efficient. The sources of allocative inefficiency at the white button mushroom production units were identified as the overuse of electricity and diesel fuel. Equipment and water for irrigation sources had the right proportions of input use by all the DMUs (zero slacks). This implies the fact that white button mushroom producers have put emphasis on optimal use of water. Optimized water use is generally due to the specific water requirement amount of mushroom; hence, farmers do their best for optimal consumption of water.

Table 5. Ranking 10 superior efficient farmers in white button mushroom production.

Rank	Farmer No.	Frequency in referent set
1	13	10
2	9	6
3	22	6
4	21	5
5	3	4
6	15	4
7	17	3
8	23	3
9	14	2
10	19	2

Table 6. Input slacks in BCC model.

Units	Efficiency score	Compost	Fuel	Chemicals	Electricity	Equipment and water for irrigation	Human labour	Output
1	1	0	0	0	0	0	0	0
2	0.94	0	0	0.37	0	0	0.05	0
3	1	0	0	0	0	0	0	0
4	0.9	0	0	0	0	0.36	0.34	0
5	0.89	0.50	0	0	0	0	0	0
6	0.91	0	0	0.50	3.08	0	0.08	0
7	0.82	0	0	0.80	4.20	0	0	0
8	1	0	0	0	0	0	0	0
9	1	0	0	0	0	0	0	0
10	0.91	0	1.12	0	3.10	0.15	0.01	0
11	1	0	0	0	0	0	0	0
12	0.93	0	34.55	1.98	29.85	0.04	0	0
13	1	0	0	0	0	0	0	0
14	1	0	0	0	0	0	0	0
15	1	0	0	0	0	0	0	0
16	1	0	0	0	0	0	0	0
17	1	0	0	0	0	0	0	0
18	0.83	0	0	1.29	3.29	0.30	0	0
19	1	0	0	0	0	0	0	0
20	0.98	0	8.24	0	11.84	0.17	0	0
21	1	0	0	0	0	0	0	0
22	1	0	0	0	0	0	0	0
23	1	0	0	0	0	0	0	0
24	0.89	0	0	0.89	0.47	0.08	0	0
25	0.98	0	0	0.08	6.64	0	0	0
26	0.87	0	4.97	0	3.46	0.17	0.05	0

On the basis of given results in **Table 6**, a single unit found with overuse of compost fertilizer. The high buildings and planting room with more stories was the potential reason for inefficient performance of compost resource use in such units. It is worth mentioning that the input orientation of analysis caused the output slacks of all units to be zero in the last column of **Table 6**.

3.2.4. Setting Realistic Input Levels for Inefficient Producers

In order to calculate the total energy input saving, the

average pure technical efficiency of inefficient farmers (0.955) was multiplied by the present use of energy inputs (133 MJ/kg) (Equation (3)). Namely, the total energy input should be reduced about 16%. **Table 7** shows the optimum energy requirement for white button mushroom production, based on the results of BCC model. Also the quantity and percentage of energy saving with respect to present use of energy is illustrated. As can be seen, optimum energy requirement for diesel fuel energy (50.89 MJ·ton⁻¹) was the highest, followed by FYM (37.32 MJ·ton⁻¹), electricity (30.34 MJ·ton⁻¹) energies.

Input	Present use (MJ·ton ⁻¹)	Target use (MJ·ton ⁻¹)	Energy saving (MJ·ton ⁻¹)	Contribution to energy saving (%)
1. Compost	41.41	37.32	4.08	9.68
2. Fossil fuel	60.85	50.89	9.95	14.5
3. Chemicals	1.18	0.57	0.62	41.92
4. Electricity	39.71	30.34	9.37	21.8
5. Water and equipment	0.78	0.60	0.18	21.21
6. Human labour	0.51	0.41	0.09	15.58
Total	144.47	120.13	24.3	15.63

Table 7. Energy optimization from different sources of energy provided following the results of this study.

The total optimum energy requirement for mushroom production was calculated to be 120.15 MJ·ton⁻¹. In the last column of Table 7 the contribution of each energy input in energy saving is presented. It is evident that 41.92% of chemicals energy, 21.8% of electricity energy and 21.21% of water for irrigation and equipment energy. which had the highest inefficiencies, could be saved. The high percentage of chemicals energy saving resulted from the low efficiency of chemical fertilizer while it is more than plant needs. There was a common belief between the farmers that increased input use especially chemical fertilizers and chemicals will increase the yield. Moreover, energy usage from human labour, machinery, fossil fuels and compost inputs could be saved by 15.58%, 14.5%, 9.68%, respectively. The results also showed that total energy input for mushroom production could be saved by 15.63% (24.3 MJ·ton⁻¹), if the recommendations of this study are followed.

Figure 2 shows the shares of the various sources from total input energy saving. It is evident that the highest contribution to the total energy saving was 42% for fossil fuel energy. This shows that in the case of fossil fuel energy, there is a greater scope to increase the efficiency of energy consumption. It was due to the relatively high contribution of fossil fuel energy from total energy consumption in present condition. Apart from fossil fuel energy, it is evident that, the contributions of compost and electricity energy inputs from total energy saving were 31% and 25%, respectively. Also, energy saving from human labour had the lowest share from total energy saving.

4. Conclusions

To sum it up, the results of DEA application suggested that there was a great potential for improving energy and economical efficiencies of edible mushroom producers in Alburz, Iran. Based on the performed analysis in this study the following results were drawn:

• Fossil fuels (53.87 MJ·ton⁻¹), compost fertilizer

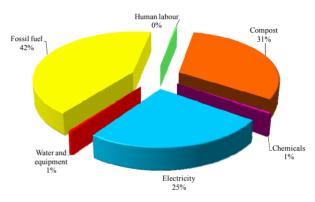


Figure 2. Distribution of energy inputs in energy saving for white button mushroom production.

(40.57 MJ·ton⁻¹) and electricity (40.57 MJ·ton⁻¹) are mostly contributed inputs to energy consumption in which fossil fuels (diesel fuel and natural gas) had the first rank (40.43%).

- From the total of 26 farmers, 14 farmers had the pure technical efficiency score (BCC model) of 1. Moreover, from the technically efficient farmers (CCR model) 12 farmers had the technical efficiency score of unity.
- The results showed that the average values of TE, PTE and SE were 0.955, 0.956 and 0.999, respectively.
- Efficient farmers No. 13, 9 and 22 with had the highest appearance times in referent set.
- The sources of allocative inefficiency (with overuse) at the white button mushroom production units were identified as electricity and diesel fuel.
- The total optimum energy requirement for mushroom production was calculated to be 120.15 MJ·ton⁻¹ in which diesel fuel energy (50.89 MJ·ton⁻¹), FYM (37.32 MJ·ton⁻¹) and electricity (30.34 MJ·ton⁻¹) energies were in the highest degree.

As a result of this study, effectiveness of use of all inputs can be achieved in an informed and efficient production system. In this regard, extension programs to-

ward the development of such systems should be taken into consideration. Here, the extension officers of agricultural institutions would be striking for managing and establishing the energy and economically efficient and environmentally sensitive edible mushroom production systems in this and other similar socio-economic regions. Moreover, providing farmers with suitable compost fertilizer can lead to highly efficient producers.

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