

Review

Meta-Analysis on Energy-Use Patterns of Cropping Systems in Iran

Morteza Zangeneh ^{1,*} , Narges Banaeian ^{1,*}  and Sean Clark ² 

¹ Department of Agricultural Mechanization Engineering, Faculty of Agricultural Sciences, University of Guilan, Rasht 4199613776, Iran

² Department of Agriculture and Natural Resources, Berea College, Berea, KY 40404, USA; clarks@berea.edu

* Correspondence: zanganeh@guilan.ac.ir (M.Z.); banaeian@guilan.ac.ir (N.B.); Tel.: +98-918-9110115 (M.Z.); +98-918-0195190 (N.B.)

Abstract: We present a meta-analysis of energy-consumption and environmental-emissions patterns in Iranian cropping systems using data collected from articles published between 2008 and 2018 for 21 different crops. The results show that the crops consuming the most energy per hectare are tomato, sugarcane, cucumber and alfalfa, while sunflower consumed the least. The average total energy input for all crops in Iran during the study period was 48,029 MJ ha⁻¹. Our analysis revealed that potato has the highest potential to reduce energy consumption and that electricity and fertilizer inputs have the most potential for energy savings in cropping systems. Not all studies reviewed addressed the factors that create energy consumption patterns and environmental emissions. Therefore, eight indicators were modeled in this meta-analysis, which include Total Energy Input, Energy Productivity, Energy Use Efficiency, Net Energy, Greenhouse Gas Emissions, Technical Efficiency, Pure Technical Efficiency and Scale Efficiency. The effects of region (which was analyzed in terms of climate), year and crop or product type on these eight indicators were modeled using meta-regression and the nonparametric Kruskal–Wallis test. To create a comprehensive picture and roadmap for future research, the process of the agricultural-systems analysis cycle is discussed. This review and meta-analysis can be used as a guide to provide useful information to researchers working on the energy dynamics of agricultural systems, especially in Iran, and in making their choices of crop types and regions in need of study.

Keywords: energy indicators; environmental assessment; greenhouse gas emissions; cropping systems; climate; sustainability



Citation: Zangeneh, M.; Banaeian, N.; Clark, S. Meta-Analysis on Energy-Use Patterns of Cropping Systems in Iran. *Sustainability* **2021**, *13*, 3868. <https://doi.org/10.3390/su13073868>

Academic Editor:
Mohammad Valipour

Received: 14 March 2021
Accepted: 30 March 2021
Published: 31 March 2021

Publisher's Note: MDPI stays neutral with regard to jurisdictional claims in published maps and institutional affiliations.



Copyright: © 2021 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (<https://creativecommons.org/licenses/by/4.0/>).

1. Introduction

In recent years, the number of publications on energy and environmental aspects of agricultural systems has increased in Iran, particularly on energy consumption patterns and environmental emissions [1]. The considerable number of publications in this field of research has necessitated a comprehensive and analytical study of the various dimensions of such studies. The method of conducting studies in this field has evolved in recent years. However, there are challenges facing researchers in this field. The first question of the present study is: Despite the wide range of indicators and methods used by previous researchers, have all aspects of cropping systems been examined or are there still issues in the systems analysis process that have been neglected? Given the great variety of production systems in the agricultural sector, have the necessary standards been met by researchers in sampling and data collection? In terms of the system analysis process, it is incomplete to study the current state of a system without intervening to improve a system. Therefore, the next question is whether, in the studies, interventions in the energy consumption pattern have been done by researchers and whether the effect of those interventions on improving the energy consumption pattern, economic model and reducing environmental emissions has been studied? Another important question is whether a hypothesis has been made about the reason for the formation of energy consumption

patterns in the products under study. Are solutions proposed to improve the current state of a system in the studies? Have social indicators been studied in studies in this field? Has the selection of the region and the product under study been based on regional and national needs and necessities? In response to the above questions, and in order to provide better directions for future research and prevent duplicate studies that have sometimes been observed in this area, it is necessary to provide researchers with sufficient information about the background of the studies. Therefore, using a systematic method, the present article provides readers with a comprehensive picture of the research background and energy consumption status in Iranian cropping systems. The sections that have been neglected in previous research are also mentioned in this article, which can be used as a basis for future research by researchers to improve the quality of studies in this field. A new procedure to supplement the shortcomings of previous studies is introduced in this article. Moreover, understanding the energy-use patterns of cropping systems is essential to addressing the sustainability challenges of agriculture. Thus, a systematic review of the relevant literature is necessary for assessing what is known and establishing future research agendas, particularly where the results of published empirical studies have generated apparently conflicting findings. A preliminary review of these studies shows that in some cases, research on the same product in different parts of the country has generated notably different results. For example, the energy efficiency index of a barley crop in Hamadan province was reported to be 2.86 [2], while in Isfahan province, it was 1.43 [3]. The existence of such differences is the rationale for this meta-analysis, which is aimed at interpreting, understanding and generalizing wherever possible based on studies published from 2008 to 2018.

Meta-analysis can be used to combine evidence from different studies into a single statistical framework, explain differences arising from conflicting findings, and identify necessary areas for future research. Iran's relatively large geographic area and climatic diversity results in a range of practices and systems for crop cultivation and leads to differences in reported values for energy and environmental indicators. The objective of study is to provide an overview of current understanding and analyze the impacts of crop type, region and year on energy and environmental indicators of cropping systems of Iran.

2. Methods

2.1. Review Protocol

This study follows PRISMA guidelines and is a companion article to one published recently on trends in methodologies and future research directions in crop energy analyses [1]. In this study, a meta-analysis was performed using published studies of energy-use patterns in cropping systems of Iran. We independently selected papers, extracted data, and assessed the results using the review process illustrated in Figure 1. Essentially, we began with the question of why the results of energy and environmental indicators for the same products differed between similar studies.

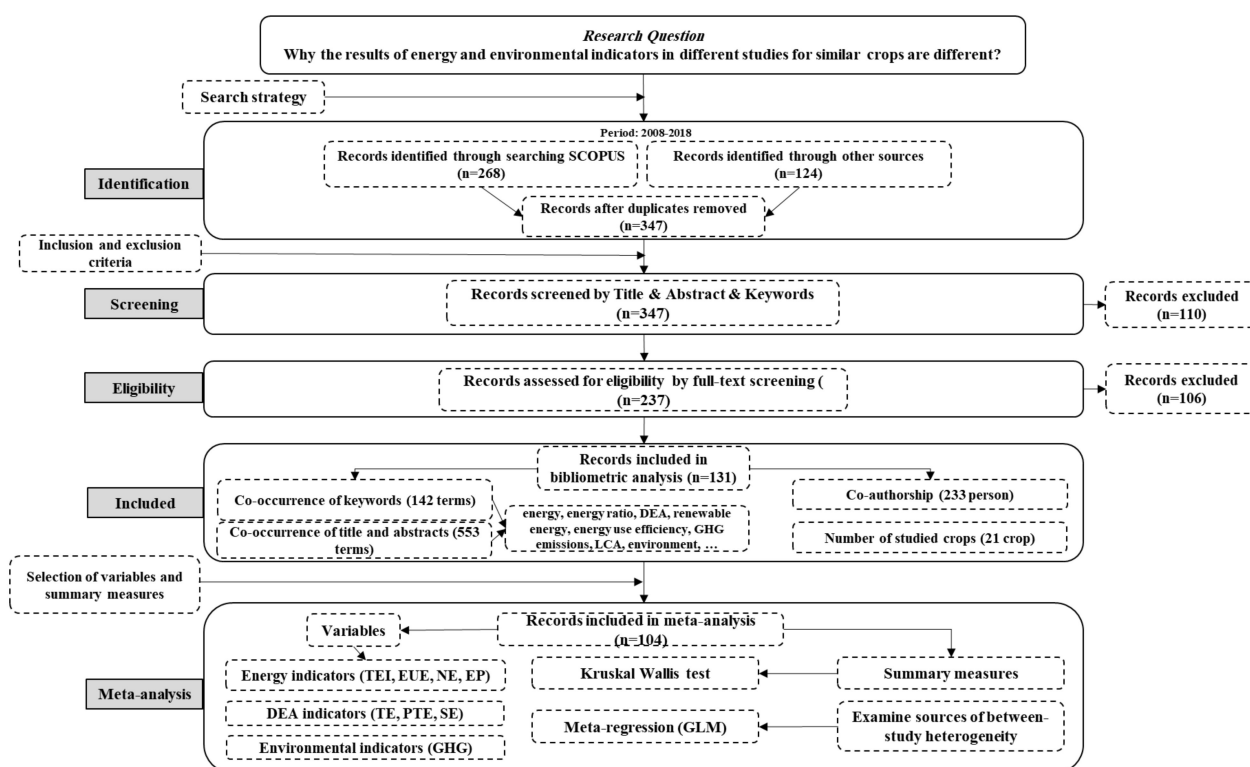


Figure 1. Schematic structure of the systematic review process.

2.2. Article Selection

The selection process included several steps described in detail in [1]. Potentially relevant titles were identified with keyword searches on SCOPUS and Google Scholar. The searches were restricted to crop studies performed in Iran, and all papers from reputable journals were given consideration. Then, abstracts and keywords were reviewed, and some inappropriate papers omitted.

2.3. Data Collection

To collect data needed for the meta-analysis, the full texts of articles were carefully reviewed, and the published results were recorded and categorized by crop or product type, study location, and year of research. Then, the values for indicators related to energy consumption patterns were extracted from the articles, including total energy input (TEI), energy use efficiency (EUE), energy productivity (EP) and net energy (NE). The type and share of each input as an amount of energy consumed were recorded. The highest t-ratio and highest marginal physical productivity (MPP) were collected as well. This initial data-extraction step was applied to all the articles reviewed.

The second step applied only to studies of cropping systems using Data Envelopment Analysis (DEA). The methodology of DEA and its application in energy analysis of cropping systems is described well by Banaeian et al. [4]. In this part, total energy savings (TES), technical efficiency (TE), pure technical efficiency (PTE) and scale efficiency (SE) were collected, as well as the shares of all inputs in the energy-saving target ratio (ESTR). The third step pertained to studies analyzing greenhouse gas (GHG) emissions and included the quantity of GHG emitted in form of CO₂ equivalents (both land-based and mass-based). In cases with more than one study of a product, the average value of an indicator was determined. For example, there were eight studies found on watermelon energy consumption [5–12], so the reported indicator values are derived from the average from all these papers.

Energy and GHG Conversion Factors

Indices used in this meta-analysis were calculated using two series of coefficients (Tables 1 and 2). Energy equivalents (energy conversion factors) are numerical values that reflect the state of energy for each input. These equivalents are used to estimate the energy indicators using collected sample data. When the amount of a consumed input is multiplied by the conversion factor, the result is equal to the energy of the input. There are some highly relevant inputs common in agricultural systems, but the energy system boundary for each is not the same in all energy studies. Table 1 shows the most relevant inputs and energy equivalents applied in energy-use-pattern studies of Iranian agricultural systems.

Table 1. Main inputs and energy equivalents used in agricultural energy systems.

Inputs (Unit)		Energy Equivalent (MJ per Unit)	References
Labor (h)		1.96	[13]
Machinery (h)		62.7	[14]
Diesel fuel (L)		56.3	[15]
Fertilizers (kg)	Nitrogen	66.14	[16]
	Phosphate	12.44	[16]
	Potassium	11.15	[3]
	Micro	120	[3]
Farmyard Manure (kg)		0.3	[17]
Chemicals (kg)	Herbicide	356.29	[18]
	Pesticide	280.44	[18]
	Fungicide	181.9	[18]
	Insecticide	101.9	[19]
Water (m ³)		1.02	[20]
Electricity (kWh)		11.93	[21]

The CO₂-eq emissions coefficients shown in Table 2 were used to calculate the amounts of the GHG emissions from crop-production inputs in Iran. The application rates of machinery, diesel fuel, farmyard manure, electricity, chemical fertilizers and biocides per hectare or per unit of product were multiplied by their corresponding emissions coefficients from Table 2.

Table 2. Greenhouse gas (GHG) emission coefficient of inputs [22].

Inputs (Unit)		GHG Coefficient (kg CO ₂ -eq Unit ^{−1})
Machinery (MJ)		0.071
Diesel fuel (L)		2.760
Fertilizers (kg)	Nitrogen (N)	1.300
	Phosphorus (P ₂ O ₅)	0.200
	Potassium (K ₂ O)	0.200
Farmyard Manure (kg)		0.126
Biocide (kg)	Herbicide	6.300
	Fungicide	5.100
	Insecticide	3.900
Electricity (kWh)		0.780

2.4. Risk of Bias in Individual Studies

A risk in studies like this one is the inability to generalize the selected sample to the entire statistical population. Researchers have often used Cochran and Neyman methods to determine the number of statistical samples. However, an important issue that researchers should consider when sampling agricultural systems is the diversity of these systems. It is preferable to select samples from all types of agricultural systems in a given area. Thus, it is necessary to first introduce the characteristics of agricultural systems in the study area and perform sampling operations based on the variety of existing systems. Failure to provide

enough information about agricultural systems and non-compliance with the proportion between the number of samples and the statistical population poses a risk to the accuracy and precision of the data as well as the values of the indicators calculated in such studies. In this regard, steps to reduce such risk in the individual studies included in this review are not possible.

2.5. Summary Measures and Synthesis

The results of this article are summarized in two sections: bibliometric analysis and meta-analysis. In the bibliometric section, the co-occurrence of keywords and co-authorship was examined using the title, abstract and keywords of the studied articles and visualized using VOSviewer software [23]. Several assumptions were made: (1) the keywords used in these texts were carefully selected by the authors; (2) the use of different keywords within a text requires some relationship between those words and; (3) the co-occurrence of keywords in the texts by different authors means that the relationships between these words are important. The meta-analysis section identifies patterns in the results of the studies included, sources of disagreement among those results, or other interesting relationships. Reasons for observed heterogeneity among the results of similar studies were interpreted using the nonparametric Kruskal–Wallis test and meta-regression. After collecting the necessary data, the results were summarized graphically. Then, for the meta-analysis, the data were analyzed using three factors: crop (product), year and regional climate.

Iran's total geographic area is 1,648,195 km², and the diversity of climates has made it possible to cultivate a variety of agricultural products across different provinces. To account for the influence of climate on cropping systems, the country was divided into 12 climate regions based on [24] (Figure 2) and each study was assigned to the appropriate climate. Table 3 lists the main climates and provinces allocated to them. It was possible for an individual province to be found within more than one climate. In such cases, the climate that occupied the largest area of a province was considered the main climate of that province and served as the basis for further analyses. Regression modeling was performed to examine the effects of crop, year and climate in each of the indicators. The relationships between each of these three factors and each of the eight indicators, which included EUE, TEI, NE, EP, GHG, TE, PTE and SE (see abbreviations), were analyzed individually using the Kruskal–Wallis test.

Table 3. Classification of Iran's provinces into different climates.

Climate	Provinces
Cold Semi Dry	East Azerbaijan, West Azerbaijan, Zanjan, Qazvin, Alborz
Very Wet and Mild	Ardabil, Guilan
Wet and Mild	Mazandaran, Golestan
Semi Dry Hot	North Khorasan, Razavi Khorasan, South Khorasan
Extreme Desert and Very Hot	Sistan and Baluchistan, Kerman
Semi Warm Desert	Isfahan, Semnan, Yazd
Very Hot Coastal Desert	Hormozgan
Warm Semi Mountains	Fars
Warm Coastal Desert	Bushehr, Khuzestan
Cold Mountains	Kohgiluyeh and Buyerahmad, Chaharmahal and Bakhtiari, Markazi, Lorestan, Kurdistan
Moderately Semi Wet	Kermanshah, Hamedan, Ilam
Temperate Desert	Qom, Tehran



Figure 2. Climates and provinces of Iran [24].

2.6. Risk of Bias across Studies

Typically, when collecting data on input consumption in crop studies, other conditions influencing the production system are assumed to be constant. However, factors such as soil conditions, climate, available technology, knowledge of producers, land-use systems, social factors, input prices, market prices, market freedom status and the extent of government intervention in market regulations (especially in Iran), among other factors, affect input consumption as well as crop production. Lack of attention to these factors has resulted in large differences in indicator values reported in different studies. The intent of this review is to interpret the heterogeneity of results in such studies based on available data.

Statistical Analysis

Regression using the General Linear Model (GLM) was performed to identify variables responsible for variations in all dependent variables, namely energy and environmental indicators. GLM usually refers to conventional linear regression models for a continuous response variable given continuous and/or categorical predictors. To determine whether the association between the response and each term in the model was statistically significant, the p -value for the term is compared to the significance level to assess the null hypothesis. The null hypothesis is that there is no association between the term and the response. Due to the nominal nature of the variables related to the cropping systems (year, climate and crop), the non-parametric Kruskal–Wallis test was used as an alternative to one-way ANOVA. Statistical significance at 0.05 and 0.01 levels was determined.

3. Results

3.1. Article Selection

A total of 131 articles were ultimately selected and used for bibliometric analysis and 104 of these were included in meta-analysis. The journals with the most published articles analyzing energy consumption patterns in Iranian agricultural systems were *Energy*, *Journal of Cleaner Production*, *Renewable and Sustainable Energy Reviews*, and *Energy Reports*, which each contained more than four articles from 2008–2018.

3.2. Study Characteristics

Quantitative and graphical summaries of word co-occurrences in the keywords, titles and abstracts among the 131 articles using VOSViewer software are shown in Table 4 and Figure 3, respectively. A threshold limit was used to plot the co-occurrence network. The minimum number of occurrences for keywords was 4 times. Accordingly, 15 keywords were found at least 4 times in articles to be related to other keywords used by researchers. For the terms used in the titles and abstracts, the threshold was set at 5 times. There were 25 terms found to meet this condition. Out of 233 authors who participated in 131 reviewed articles, 18 of them contributed more than 5 times to articles. How these terms relate to each other is shown in Figure 3.

The following terms were the most common in the articles: energy, greenhouse gas emissions, data envelopment analysis, sensitivity analysis, economic analysis, energy efficiency and technical efficiency. Shahin Rafiee, a faculty member at the University of Tehran, has the most joint publications, co-authoring 43 papers between 2008 and 2018. There were 18 authors found to have published more than 5 articles.

Table 4. Bibliometric analysis of the selected articles.

Description	Categories of Co-Occurrences		
	Keywords	Titles and Abstracts	Authors
Total items	142	553	233
Minimum number of occurrences	4	5	5
Number of items meet the threshold	15	25	18

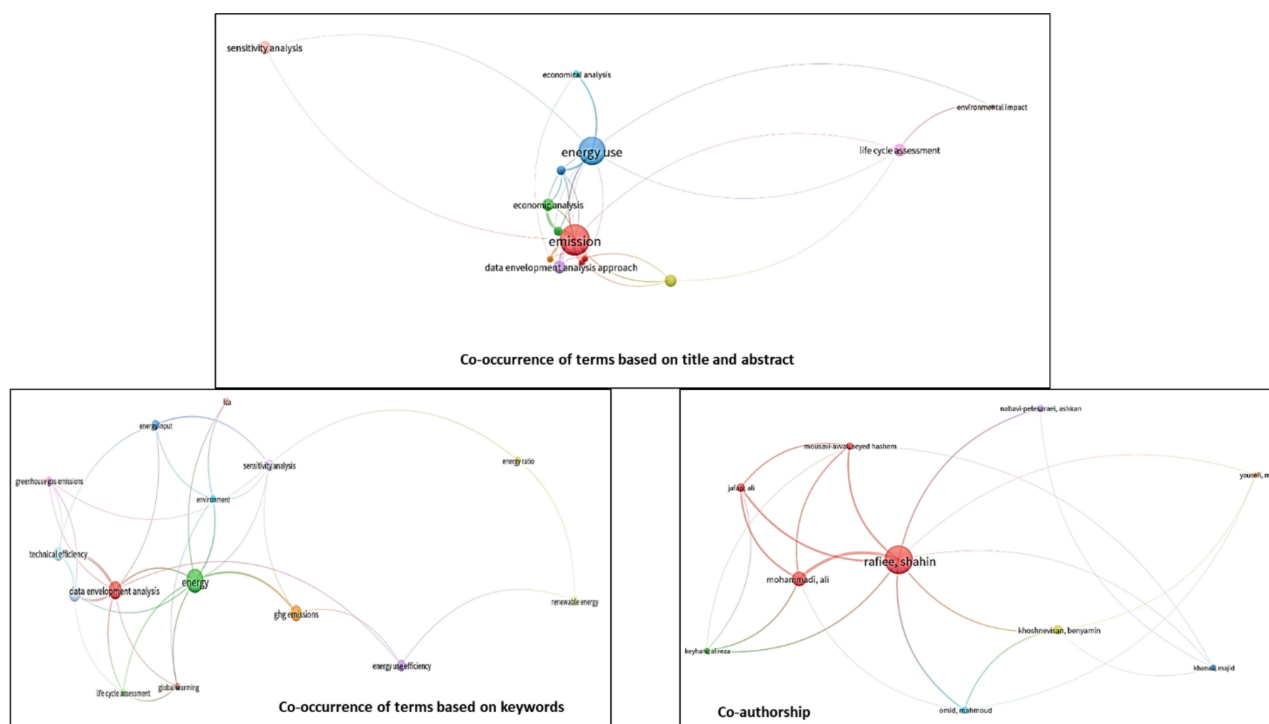


Figure 3. Graphical display of the co-occurrence analysis.

3.3. Synthesis of Results

To create a useful overview of energy consumption in Iran's cropping systems, we first present a summary of the most important results published on input consumption, environmental impacts, and data envelopment analysis. Then, the effects of the independent variables of year, climate and crop on the dependent variables of energy and environmental indicators are analyzed and modeled. Finally, the relationship between each of the independent and dependent variables is analyzed using the non-parametric Kruskal–Wallis test.

3.3.1. Share of Inputs

A percentage breakdown of the TEI into input categories for each crop is illustrated in Figure 4. For crops in which there was more than one article, averages are used. Fertilizer and fuel are the main consumers of energy in many crop systems. Authors often report the share of important energy-consuming inputs and make recommendations for improvements to reduce the TEI. For example, Hosseinzadeh-Bandbafha et al. [25] recommended that to benefit from a balanced use of fertilizers, emphasis should be placed on using high-quality seeds, optimal timing, precisely timed irrigations and better agronomic practices in farm operations (e.g., reduce field passes and using shallow tillage) for peanut farms at the north of Iran.

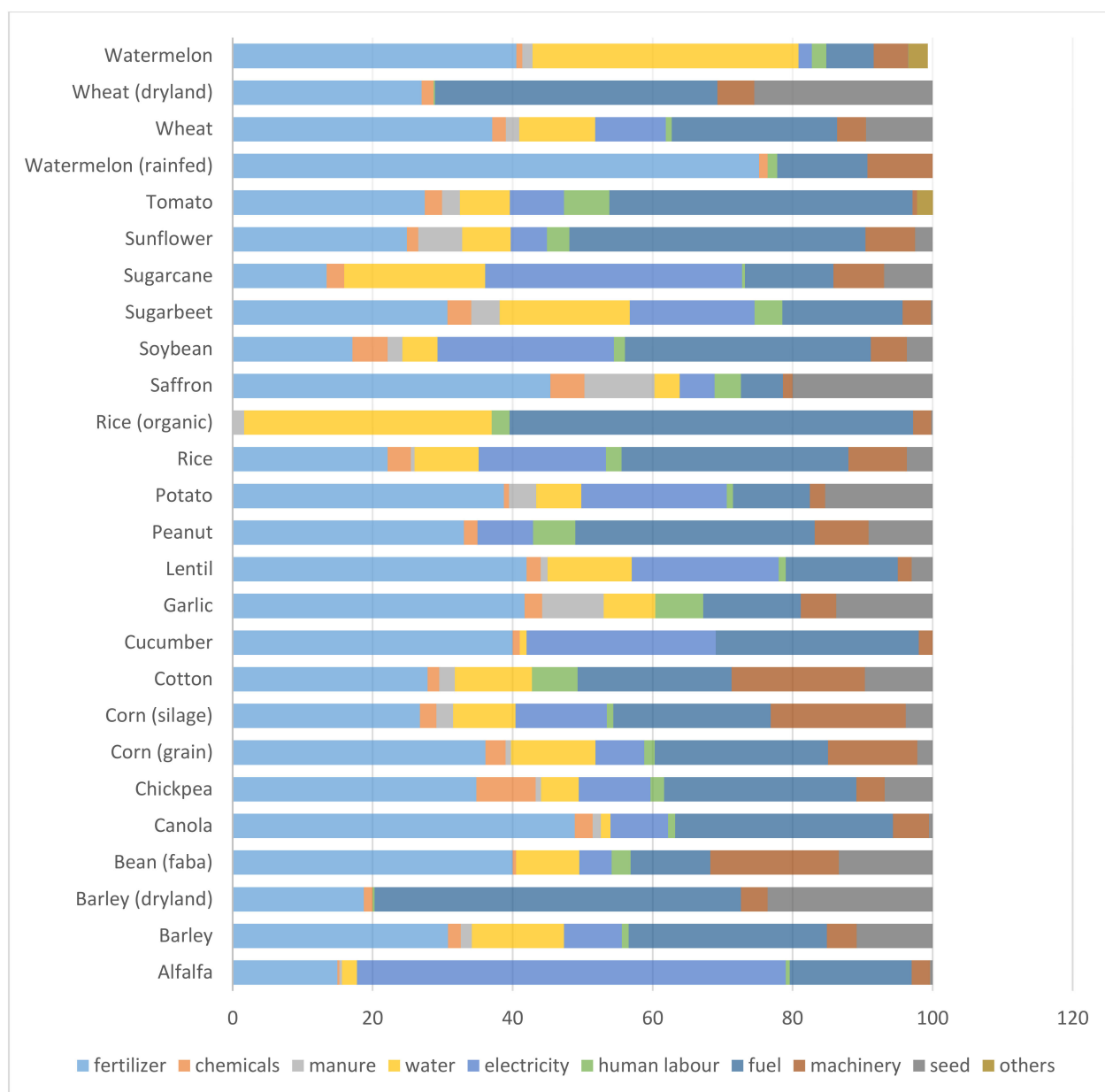


Figure 4. Share of inputs (%) in total energy input (TEI).

3.3.2. Environmental Impact

In addition to examining energy and economic indices, many of the articles we found addressed environmental issues of cropping systems. Here, we focus first on the 54 studies that investigated the environmental impacts associated with energy use. Production, transportation, storage, input distribution and machinery use consume fossil fuels and other energy sources that emit GHGs into the atmosphere. The most common method of calculating environmental impacts uses emission factors (Table 2) that are multiplied by the corresponding input inventory, such as diesel fuel, chemical fertilizer, biocides and irrigation water. The emissions are presented in kilograms of carbon equivalent for different farming activities and can be compared to alternatives, such as biofuels and renewable energy sources [26].

Among the reviewed papers, GHG emissions and global warming potential (GWP) are sometimes calculated manually using emission factors (CO₂ equivalents) or determined using a life cycle inventory (LCI) and software analysis. The latter allows a wide range of

environmental impacts to be assessed. Khoshnevisan et al. [27], for example, used Ecoinvent LCI and SimaPro Software for life cycle assessment (LCA) of rice production systems in northern Iran. Several authors have subsequently performed comprehensive studies of individual cropping systems with LCA [28,29]. Further, the energy and environmental impacts of different cropping systems have been compared using LCA [30,31].

Table 5 summarizes GHG emissions reported for crops studied in Iran. Researchers have applied emission factors to report GHG emissions of crops, including potato [26], cotton [32], wheat [33] and barley [34]. As indicated in Table 5, GHG emissions of some crops have been determined and reported in multiple studies. In such cases, the average is presented and is likely more reliable. It is notable that there is a wide range of CO₂ emissions reported across these crop production systems. For example, 181,190 and 558 kg CO₂-eq were produced per ha of alfalfa [35] and potato [36] production, respectively. In most cases, reported values for GHG emission in Iranian studies are similar to those of other studies around the world. For instance, Yousefi et al. [37] reported 2994.66 kg CO₂-eq per ha for corn production, while Camargo et al. [38] reported that the GHG emissions associated with corn production ranged from 2441 to 4201 kg CO₂-eq per ha per year in several publications. However, there are some papers with results differing significantly with those reported for GHG emissions for agricultural systems outside of Iran in crops like alfalfa and corn [39,40].

Table 5. The amounts of GHG emissions reported for farm crops of Iran.

Crop	GHG Emission		References
	kg CO ₂ t ⁻¹ (Mass Base)	kg CO ₂ ha ⁻¹ (Land Base)	
Alfalfa	52.09	181,190	[35,41]
Barley	-	628	[34,42]
Canola	-	836.6	[28,42,43]
Chickpea	3032.6	6884.14	[44]
Corn	-	2994.66	[37]
Cotton	-	1195.25	[32]
Lentil	3593.2	7259.31	[45]
Peanut	311.19	697	[25,46,47]
Potato	116.4	558	[26,36,48]
Rice	1101	3197.00	[27,42,49–51]
Saffron	-	6545.8	[52]
Soybean	455,515	1197	[42,53,54]
Sugar beet	-	9847.77	[55]
Sugarcane	-	8249.12	[56]
Tomato	200	-	[29]
Watermelon	-	5299	[6,7,9,11,12]
Wheat	1600	2155	[6,31,33,42,57–64]
Grape	508.63	-	[65]
Tobacco	1883.90	3638.98	[66]

3.3.3. Data Envelopment Analysis

A summary of findings from DEA studies is presented in Figure 5, including estimated TE, PTE, and SE of crops, TES, and the share of each input in the total energy-saving ratio. As shown in the share of agricultural inputs in the total energy-saving ratio, electricity and fertilizer inputs have the most potential for energy reductions across the various crops. Potato has the highest potential to reduce energy consumption. By contrast, there is much less potential for peanut. TES in peanut farms of Guilan province were reported to be 9.15%, with machinery accounting for the largest fraction at 13.93% of the total, followed by biocides (10.23%) and chemical fertilizers (9.79%) [25].

Examples of complementary analysis with the DEA method include super-efficiency analysis by Mohammadi, Rafiee, Jafari, Dalgaard, Knudsen, Keyhani, Mousavi-Avval and Hermansen [55] and fractional regression model by Raheli et al. [67]. Mohammadi, Rafiee,

Jafari, Dalgaard, Knudsen, Keyhani, Mousavi-Avval and Hermansen [55] first applied LCA and DEA methodologies to soybean farming to benchmark the level of operational input efficiency and potential GHG emissions reductions by farmers. In addition, DEA has been widely used in combination with CO₂-eq emissions data to determine the potential of GHG emissions reductions in agricultural systems [8,25,68].

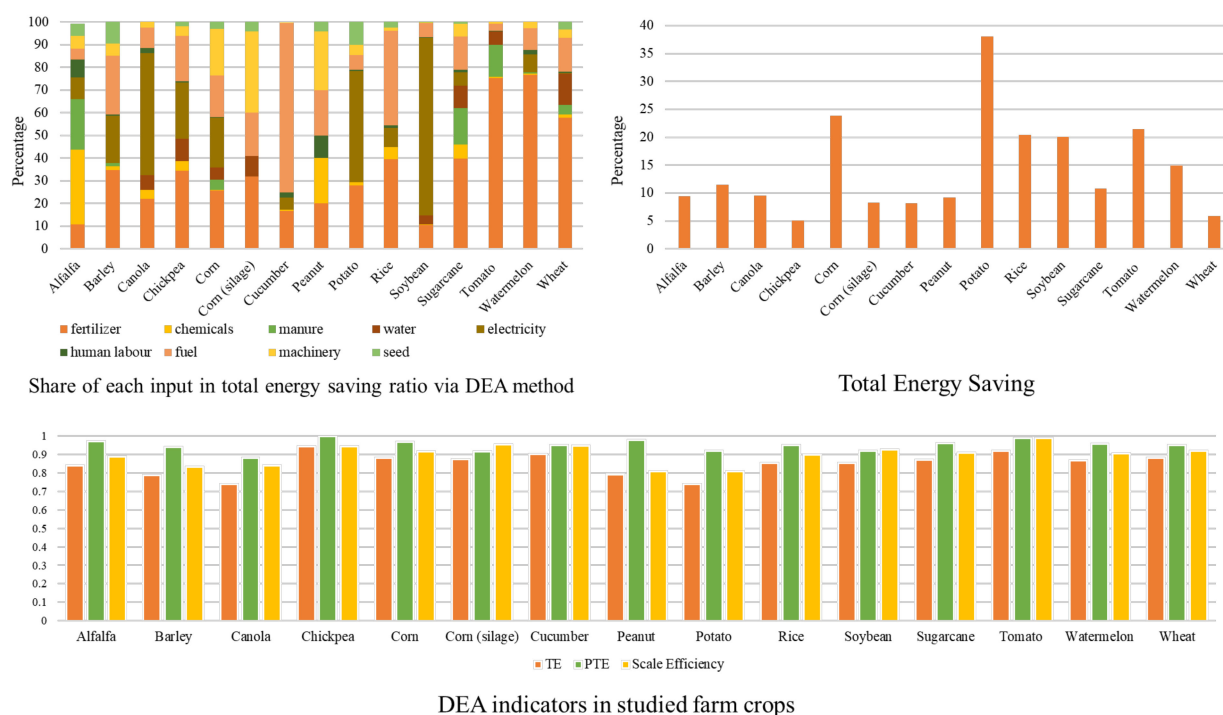


Figure 5. Summary of results from data envelopment analyses (DEA).

3.3.4. Economic Indicators

As the field of energy studies flourished in Iran, some researchers began adding economic indicators to their studies. Some studies reported costs of inputs in detail [69], while others only reported economic indicators. Various scales and units have been applied in reporting economic indices. For example, Ghorbani et al. [70] reported Total Production Cost, Gross Return and Net Return in \$ ha⁻¹, \$ kg⁻¹ and \$ MJ⁻¹, respectively. Such economic analyses were common in studies from 2010 to 2012. Then, research interest gradually shifted to other techniques. Just two papers in 2013 and 2014 applied economic analysis [71,72], and none did in 2015. In 2016, Sahabi, Feizi and Karbasi [20] compared energy and economic aspects of saffron and wheat in northern Iran. In 2017, Mousavi-Avval et al. [73] used the same economic indicators to analyze canola production in northern Iran. In recent years, studies have become more varied, and new analyses have been added with environmental impacts highlighted.

3.3.5. GLM Results

In our review, we identified 21 different product types or crops among the articles to be used for modeling. We created eight linear regression models with the following energy and environmental indicators as dependent variables: GHG, SE, PTE, TE, NE, TEI, EP and EUE. The results of these models are presented in Table 6. We used three independent variables for regression modeling including region, year and crop. The “region” represents the climates of Iran, which includes 12 climates (Table 3). Four of the eight models predicted the dependent variable using the terms provided.

Of course, differences among regions in energy consumption and environmental emissions for the same product are not simply due to climate differences. There are other

differences, such as the level of technology used, planting methods, harvesting and storing, farmer knowledge, access to farming inputs, etc. However, because the articles reviewed did not always report the specific characteristics of cropping systems, we are limited in explaining the differences in energy and environmental indicators between different regions. The “year” refers to when the study was performed as reported in the article. The “crop” factor was influential in the regression models, which seems obvious because the amounts and types of inputs consumed and the production outputs of various products are fundamentally different from one to another. Thus, crop type would be expected to be a very important factor.

Table 6. General linear model results.

Dependent Variables		Source						R Squared
		Corrected Model	Intercept	Year ¹	Region	Crop	Region * Crop	
GHG	F	5.668 **	2.233	2.256	9.003 **	5.433 **	8.257 **	0.886
	Sig.	0.000	0.155	0.153	0.000	0.001	0.003	
SE	F	3.143	3.882	3.865	0.751	3.995	.	0.962
	Sig.	0.268	0.188	0.188	0.571	0.216	.	
PTE	F	5.535	0.067	0.113	12.696	4.948	.	0.978
	Sig.	0.164	0.820	0.769	0.073	0.179	.	
TE	F	3.838	0.707	0.579	1.118	4.847	.	0.953
	Sig.	0.147	0.462	0.502	0.434	0.111	.	
NE	F	23.618 **	2.568	2.526	5.375 **	47.819 **	7.735 **	0.955
	Sig.	0.000	0.114	0.117	0.000	0.000	0.000	
TEI	F	3.997 **	0.064	0.073	1.56	5.355 **	1.556	0.766
	Sig.	0.000	0.801	0.787	0.137	0.000	0.071	
EP	F	0.284	0.394	0.396	0.329	0.237	0.232	0.201
	Sig.	1.000	0.532	0.531	0.97	1.000	1.000	
EUE	F	13.797 **	0.917	0.872	4.099 **	31.49 **	2.612 **	0.923
	Sig.	0.000	0.342	0.354	0.000	0.000	0.001	

** Significant at the 0.01 level (2-tailed). * Significant at the 0.05 level (2-tailed). ¹ Year is covariate for all dependent variables.

In the model developed for GHG, the variables of region, crop and their interaction (crop*region) have significant effects. Therefore, there are significant differences between crops. Applications of different inputs, including chemicals and diesel fuel in field operations, can cause differences in GHG emissions. Agricultural activities inevitably result in emissions of all three greenhouse gases: carbon dioxide (CO₂), methane (CH₄) and nitrous oxide (N₂O). Methane is mainly released through farm manure; nitrous oxide is produced from the combustion of fossil fuels, manure, soil cultivation and decomposition of crop residue; and carbon dioxide is generated and released into the atmosphere through fossil fuels and the decomposition of crop residue.

The regression analysis indicates that GHG emissions differ across the regions of Iran, meaning that different climates can cause different GHG emissions. Climate is shaped by short- and long-term characters, including temperature, average rainfall, floods, drought, etc. All these climate characteristics impact food quality, distribution and access patterns [74]. Alboghdady and El-Hendawy Salah [75] used panel data from 20 countries in the Middle East and North Africa from 1961 to 2009 to assess the impact of climate change and variability on agricultural production. The results showed that a 1% increase in winter temperature results in a 1.12% decrease in agricultural production. Our analysis indicates that warmer and drier climates have more emissions. This may be due to the lack of optimal conditions for production and a resulting need to consume more inputs to compensate.

A more detailed comparison of wheat production across four climates is shown in Figures 6 and 7. In the semi-warm climate, where Isfahan province is located, five studies have been performed on GHG emissions from wheat production [33,58,59,76,77]. They

show that the consumption of livestock manure, chemical fertilizers and electricity is much higher than in other climates of Iran. It seems that the dry climate of Isfahan province increases the need to pump water from agricultural wells using electricity. Thus, emissions from electrical inputs are higher than in other climates.

On the other hand, Golestan province, which has a wet and mild climate, has the lowest electricity consumption. Data show that the amount of fuel consumed in Kermanshah province, located in a semi-arid climate, is much higher than other climates. Kermanshah farms also need irrigation from well sources, mainly due to drought. If the wells in this province are not electrified, higher fuel consumption results from pumping water. In Khuzestan province electricity consumption has not been reported, likely due to the abundance of surface water to irrigate farms.

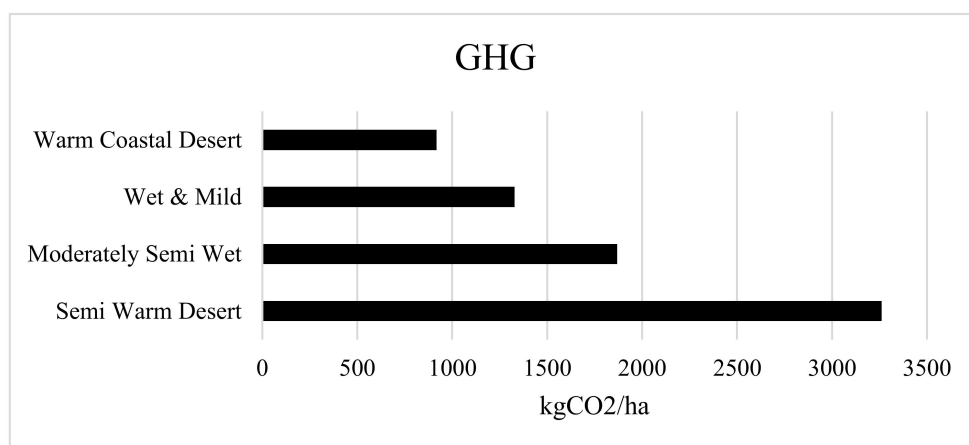


Figure 6. GHG emissions quantity for wheat in different climates.

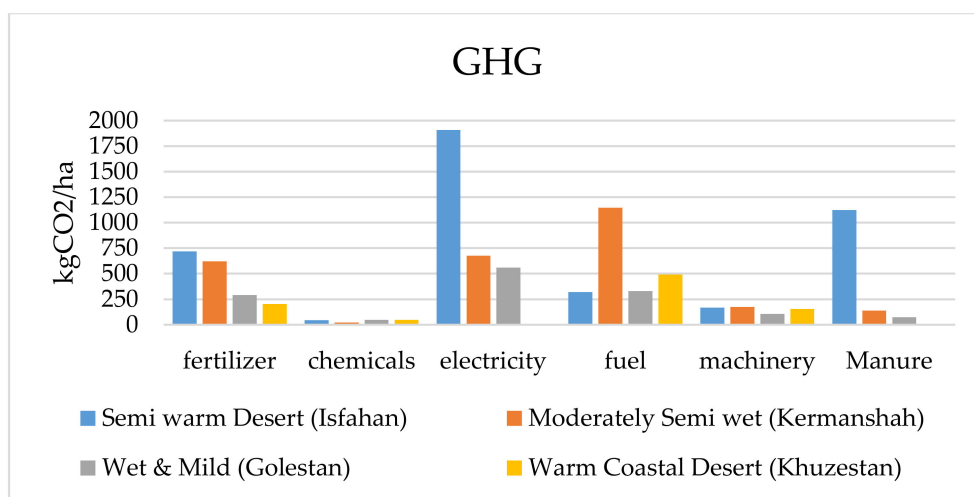


Figure 7. GHG quantity of inputs in different climates for wheat.

Examination of the NE regression model shows that, like the model for GHG emissions, the variables of region, crop and interactions of these two factors have significant effects. The same is true for the EUE model. However, for the input energy indicator, only the crop variable has a significant effect. Our four regression models show that climate factors, product type and which products are produced in which climate can predict GHG emissions, NE and EUE.

Our review revealed that from 2008 to 2018, ten studies were conducted on the energy consumption patterns of potatoes in Iran. Most of these studies took place in the main centers of potato production, namely Ardabil, Hamedan and Isfahan provinces. Comparing

the EUE in these areas indicates that Ardabil province has the best conditions (Figure 8). After Ardabil are Isfahan, Hamedan and Tehran provinces.

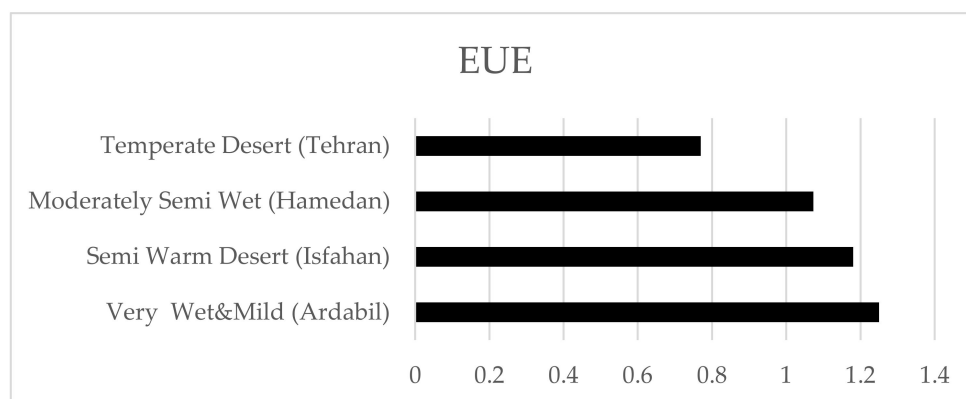


Figure 8. Energy use efficiency of potato production in different climates of Iran.

3.3.6. Production Function and Sensitivity Analysis

After calculating energy inputs, some researchers have sought to analyze relationships between energy inputs and crop output using the Cobb–Douglass production function [78,79]. The results of these studies are summarized in Table 7. Here, the highest t-ratio belongs to the most effective input influencing crop yield. A sensitivity analysis identifies how different values of an independent variable (energy input) can affect a specific dependent variable (crop yield). Sensitivity analyses were first used by Rafiee et al. [80] and Mobtaker, Keyhani, Mohammadi, Rafiee and Akram [2]. Mousavi-Avval et al. [81] reported the highest MPP index for fertilizer, water and machinery energy inputs on canola production. They showed that a 1 MJ overuse of each of these energy inputs will increase the canola production by 0.61, 0.24 and 0.93 kg, respectively. Generally, as can be seen from the results of sensitivity analyses, fuels, machines and chemical fertilizers had the greatest impact on crop yield.

Table 7. Highest t-ratio and MPP of crops produced in Iran.

Crop	Highest t-Ratio	Highest MPP	References
Alfalfa	Diesel fuel	Diesel fuel	[82]
Alfalfa	Machinery	Machinery	[83]
Bean (red)	Machinery	Machinery	[84]
Canola	Nitrogen	-	[43]
Canola	Seed	-	[73]
Corn (silage)	Water	Seed	[40]
Corn (silage)	Biocide	Biocide	[3]
Cucumber	Chemicals and diesel fuel	Human labor	[85]
Cucumber	Machinery	-	[86]
Potato	Seed	-	[87]
Potato	Water	Seed	[88]
Potato	-	Fertilizer	[26]
Rice	Fuel	Fuel	[89]
Rice	Chemicals	-	[90]
Rice	Fertilizers	Fertilizers	[51]
Rice	Manure	-	[91]

Table 7. Cont.

Crop	Highest t-Ratio	Highest MPP	References
Soybean	Fertilizers	Machinery	[92]
Soybean	Seed	Seed	[93]
Soybean	Human labor	-	[54]
Sunflower	Diesel fuel	Chemicals	[21]
Watermelon	Human labor	Human labor	[11]
Watermelon	Chemicals	-	[12]
Wheat	Machinery	Machinery	[18]
Wheat	-	Fertilizers	[59]
Wheat (dryland)	-	Water	[94]

3.4. Risk of Bias across Studies

Non-Parametric Test Results

Relationships between the three factors (region, year and crop) and each of the eight indicators (EUE, TEI, NE, EP, GHG, TE, PTE and SE) were tested using the Kruskal–Wallis test, the non-parametric equivalent of a one-way ANOVA. The results are summarized in Table 8. We can conclude that the regions have different NE, TEI and EUE. The crops also differ in terms of NE, TEI, EUE and EP (Figure 9). No significant relationship was found between the year variable and any of the energy and environmental indicators. This suggests that over the 10-year period, cropping systems were stable in terms of energy consumption patterns and environmental emissions. For comparison, Han and Wu [95] explored the impacts of changes in China's agricultural structure on factors such as energy intensity of agricultural production. Their results showed that the results of six vegetable production regions show great regional heterogeneity, which is mainly due to the scale economy effects and incremental increases in mechanization.

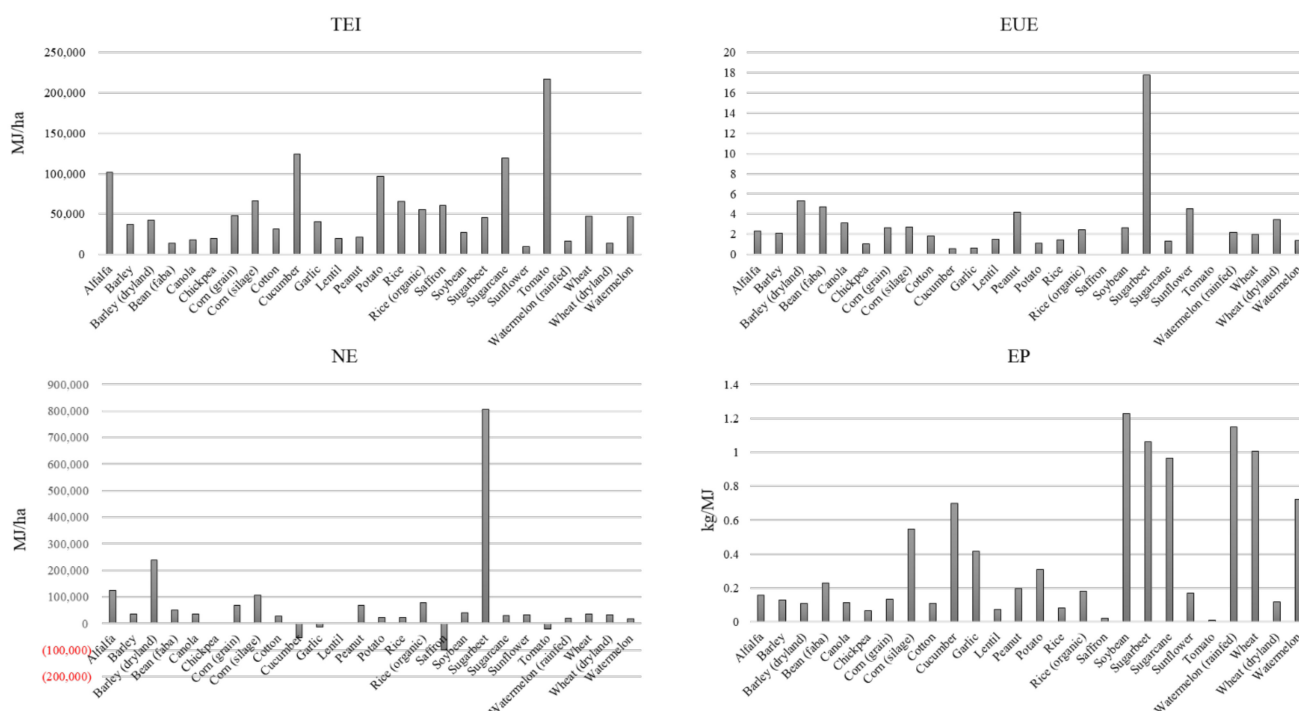


Figure 9. Average of energy indicators for different crops.

Table 8. Kruskal–Wallis test results.

Test Variables	Grouping Variables					
	Region		Year		Crop	
	Chi-Square	Asymp. Sig.	Chi-Square	Asymp. Sig.	Chi-Square	Asymp. Sig.
GHG	7.254	0.403	3.595	0.731	16.340	0.231
SE	5.003	0.543	5.523	0.479	16.463	0.225
PTE	8.849	0.182	5.995	0.420	11.937	0.533
TE	2.322	0.888	3.107	0.795	13.216	0.431
NE	27.836 **	0.002	15.461	0.079	54.399 **	0.000
TEI	29.619 **	0.001	13.084	0.219	75.437 **	0.000
EP	13.292	0.208	7.453	0.682	59.404 **	0.000
EUE	30.082 **	0.001	11.793	0.299	61.791 **	0.000

** Significant at the 0.01 level (2-tailed).

4. Discussion

Agricultural systems are often described in terms of their diversity, or different types of systems, and their heterogeneity, or variation in the physical, biological, and human components within each type of system. The diversity and complexity of agricultural systems throughout the world can be tracked and evaluated with a variety of economic, environmental, and social performance indicators [96]. Our review revealed that in some published papers, values for energy and environmental indicators are reported, but no explanations are given for the causes or reasons for the status in these indicators such as Amanloo and Mobtaker [39,97,98]. In some cases, results have been compared with similar studies. For example, Mohammadi-Barsari, Firouzi and Aminpanah [12] stated that calculated energy productivity is higher for watermelon farms with advanced technology (0.66 kg MJ^{-1}) than for farms with less advanced technology (0.59 kg MJ^{-1}) in the semi-arid region of Hamadan province. Agricultural systems can vary in many aspects, such as farm size, seed type, date of cultivation, irrigation, tillage, harvest methods, etc., and some researchers have investigated these factors from the viewpoint of energy use, economics and other indicators, as shown in Table 9. Most of these studies occurred in the years 2011 and 2012 and often applied ANOVA and the Duncan multiple range test to compare means of classified groups. Statistical design has also been implemented to audit energy consumption and identify the share of energy inputs for different tillage systems [99].

Table 9. Energy indices in different cropping systems of Iran.

Crop	Region	Classification	IE (MJ ha^{-1})	EUE	Reference
Potato	Ardabil	Farm size	0.1–2 ha	39,677.9	[13]
			2.1–5 ha	37,908.7	
			5 < ha	37,482.5	
Red bean	Kurdistan	Farm size	0.1	105,540.2	[84]
			0.2	47,571	
			0.5	43,725.4	
Lentil	Lorestan	Production method	Organic	5062	[100]
			Conventional	6196.5	
Soybean	Golestan	Irrigation system	Canal irrigation	17,255.96	[101]
			Pump irrigation	38,266.71	

Table 9. Cont.

Crop	Region	Classification		IE (MJ ha ⁻¹)	EUE	Reference
Potato	Hamedan	Technology level	High level	153,071.4	1.14	[87]
			Low level	157,151.12	0.95	
Rice	Mazandaran	Mechanization status	Traditional	67,356.28	3	[102]
			Semi-mechanized	67,217.95	3.08	
Alfalfa	Hamedan	Irrigation system	Traditional	821,615.19	1.82	[103]
			Modern	723,254.38	2.06	
Wheat	North Khorasan	Irrigation method	Irrigated	45,367.63	1.44	[70]
			Dryland	9354.2	3.38	
Corn silage	Alborz	Farm size	5 > ha	86,679	1.72	[40,104]
			5–10 ha	65,845	2.29	
			>10 ha	54,499	2.8	
Rice	Guilan	Seed type	Hashemi	37,155.213	1.582	[89]
			Khazar	41,332.513	1.956	
			Hybrid	44,848.813	2.458	
Rice	Guilan	Farm size	0.5 > ha	41,140	1.44	[105]
			0.5–1 ha	40,433	1.47	
			>1 ha	36,428	1.69	
Corn	Fars	Region	Seyedan	41,631.97	2.6	[106]
			Houmeh	44,730.15	2.38	
			Pasargad	38,866.64	2.88	
Rice	Mazandaran and Isfahan	Production method	Organic	134,851.6	2.43	[107]
			Conventional	155,762.7	1.11	
Canola	Khuzestan	Irrigation system	Irrigated	28,944.65	1.28	[108]
			Dryland	18,557.72	0.81	
Potato	Isfahan	Farm size	<1 ha	51,460	1.3	[26]
			1–5 ha	45,710	1.75	
			>5 ha	43,874	2.08	
Corn	Alborz	Harvesting system	Combine harvesting	49,303	5.15	[104]
			Plot harvester	49,448	4.4	
			Two stage harvesting	54,471	4.78	
Corn silage	Tehran	Farm size	<10 ha	38,841.5	3.11	[109]
			10–20 ha	36,140.3	3.47	
			20–30 ha	35,861.1	3.56	
			>30 ha	35,211.6	3.82	
Rice	Mazandaran	Traditional production method	Average	71,092.26	1.72	[110]
			Native	60,187.41	1.33	
			High yield	73,220.42	1.74	
			Hybrid	79,908.94	2.01	

Table 9. Cont.

Crop	Region	Classification	IE (MJ ha ⁻¹)	EUE	Reference
Rice	Mazandaran	Mechanized production method	Average	79,460.33	[110]
			Native	69,181.23	
			High yield	82,005.42	
			Hybrid	87,186.06	
Wheat	Isfahan	Farm size	Small	80,400	[57]
			Medium	79,290	
			Large	81,110	
Rice	Guilan	Land management	Traditional	74,200	[27]
			Consolidate	57,000	
Rice	Khuzestan	Planting method	Transplanting	50,022	[111]
			Direct seeding	34,543	
Watermelon	Khorasan and Semnan	Irrigation system	Full irrigation	25,626	[7]
			Reduced irrigation	3129.3	
Soybean	Golestan	Mechanization status	Modern mechanized	29,532	[54]
			Mechanized	29,599	
			Conventional (more tillage)	15,369	
			Conventional	14,657	

Some researchers reported that larger-scale farms perform better according to energy indicators [40,105]. To investigate the effect of farm size and crop type on EUE, a regression model was developed using meta-data (Table 10). The regression model indicates that the size of the farm has no significant effect on energy efficiency, but the crop type does affect the EUE.

Table 10. General linear model result for energy use efficiency (EUE).

Dependent Variables		Corrected Model	Intercept	Crop Type	Farm Size	Crop Type Farm Size	R Squared
EUE	F Sig.	1.827 0.236	49.828 0.000	5.941 ** 0.028	0.417 0.677	0.052 1.000	0.81

** Significant at the 0.01 level (2-tailed).

For comparison, Ito [112] measured regional differences in agricultural productivity in China to test the validity of a hypothesis related to agricultural technology. Qiang et al. [113] compared the agricultural disasters in the north and south of China, and the results showed that the losses in the north increased by about 0.6% every ten years, close to twice that in the south of China. In addition, agriculture in northern China was more sensitive to precipitation change, while agriculture in southern China was more sensitive to temperature change. Other papers have investigated the effects of production parameters on the energy-use patterns. For example, Banaeian and Namdari [5] investigated how farming technology and ownership of machinery, tractors and land can affect energy-use patterns in watermelon. In analyzing the heterogeneity of agricultural technology, Fei and Lin [114] used meta-frontier DEA to measure agricultural energy efficiency. The results showed that energy efficiency in the eastern region of China was significantly higher than that of the western region. Based on the provincial panel data of 1995–2014, Diao et al. [115] analyzed the agricultural productivity and its regional differences in China.

Expanding cropland onto areas under natural ecosystems reduces carbon stocks in natural vegetation and soils, with the amount of carbon released and crop yields differing markedly between temperate regions and the tropics [116]. Cassman [117] indicates that precise management and improvements in soil quality are needed to achieve high yields without causing environmental damage. No-tillage has revolutionized agricultural systems because it allows individual producers to manage larger amounts of land with fewer inputs of energy, labor and machinery [118]. Lal [119] points out that not all conservation agriculture practices, and other resource conservation technologies are applicable across all farming systems. Sustainable land management practices such as crop diversification and chemical management are potentially efficient measures that lead to environmental and economic benefits [120]. In addition, synthetic N fertilizer is the major source of GHG emissions from crop production [121]. Better management in the use of chemical fertilizers on farms can be effective in improving productivity as well as protecting water resources and minimizing greenhouse gas emissions. Therefore, creating GHG profiles of different crops in Iran can greatly help in developing and improving national mitigation plans/policies that currently lack details pertaining to the agricultural sector.

In our analysis, as the results of the Kruskal–Wallis test showed, Iran’s cropping systems have been almost constant over the 10-year period in terms of energy consumption patterns and environmental emissions based on the 21 crops evaluated. This apparent lack of improvement points to the need for researchers to identify methods to improve energy consumption patterns. Closer examinations of the characteristics of cropping systems should help researchers to identify and better understand energy consumption patterns. Differences in the indices for energy consumption and environmental emissions for similar products in different regions and years are not easily explained and will therefore require access to more precise information about the characteristics of agricultural systems.

As researchers continue in their efforts, they should pay attention to the following three points: (1) it is essential to understand the reasons for the current situation in order to devise reasonable, practical strategies for improvement; (2) farmers must be intimately involved in diagnosis of the problems and in devising improvement strategies; and (3) agricultural technologies and policies (and support systems) are complementary means of improving agricultural productivity and sustainability. If researchers can implement solutions to improve patterns of energy consumption on sample farms, they can serve as models for farmers. Agricultural systems, like any other system, need continuous analysis and improvement, so researchers must identify inputs with non-optimal consumption, design solutions to address the issues, and implement pilot projects to demonstrate improved performance.

Jones et al. [122] studied and summarized the history of agricultural systems modeling and identified lessons learned that can help guide the design and development of tools and methods for the next generation of agricultural systems. They emphasized that there are two broad categories that motivate agricultural model development: scientific understanding and decision/policy support. The complex and interacting dimensions of modeling agricultural systems are illustrated in Figure 10.

Agricultural systems are affected by many external drivers, which can be divided into four groups: technical, economic, environmental and social. These systems can also be classified and studied at various scales, such as field, farm, regional, national and global. Users include farmers, suppliers, agencies, governments and international companies, and they pursue different goals. At the field level, input consumption management is key, while at the farm level it is production optimization and business management. Shifting to the regional level, natural resource conservation, economic planning, and environmental and landscape management become the focus. At the national level, the priorities become trade policies, poverty alleviation and strategic planning, and globally, they are global business, climate change adaptation and food security to reliably feed more than 9 billion people. To improve the performance of agricultural systems, modeling and analysis at different scales

is necessary. The results of our review show that studies conducted in Iranian agricultural systems have been focused on field and farm scales.

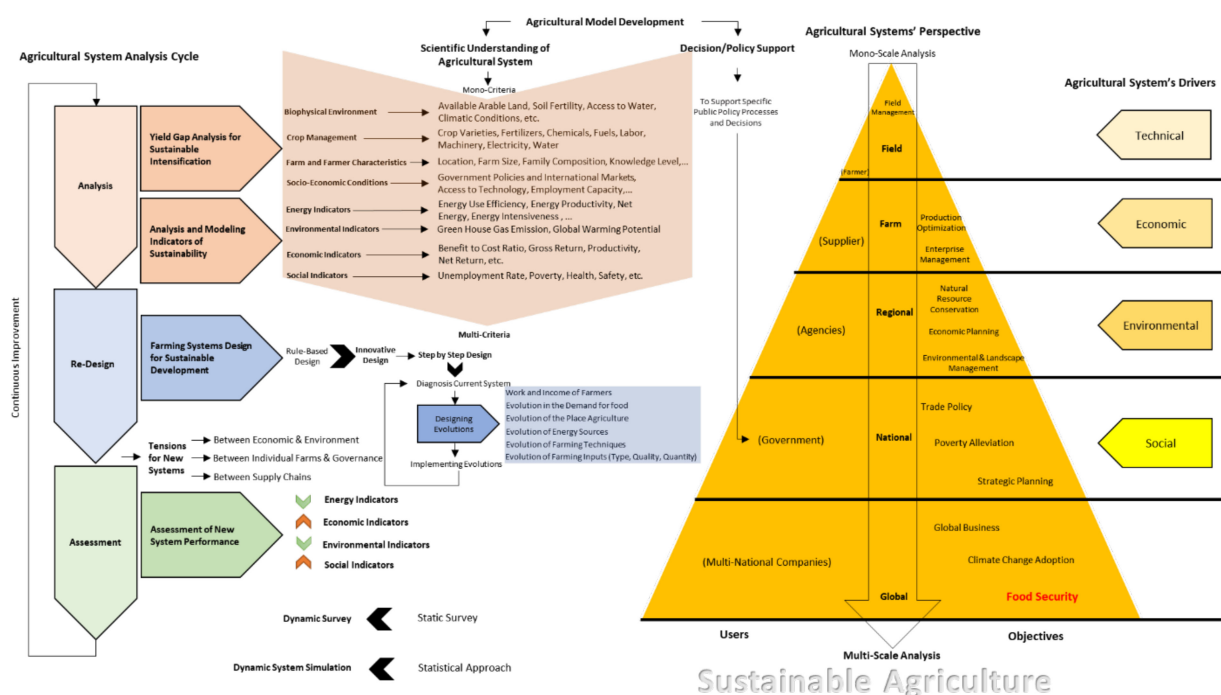


Figure 10. Dimensions and requirements of modeling and analysis of agricultural systems.

Analysis and modeling of agricultural systems should exist within a continuous cycle of at least three main phases: system analysis, system redesign and system assessment. System analysis consists of two steps: (1) yield gap analysis for sustainable intensification and (2) analysis of and modeling indicators of sustainability. Since many different factors affect these systems, it is necessary to shift the analysis and modeling from a single criterion to multiple criteria. At the field or farm level, biophysical models are used to analyze responses similar to the way that experiments on the real systems would be analyzed. Biophysical environment indicators include available arable land, soil fertility, access to water, climatic conditions, etc. Crop management indicators are also used at field and farm scales and consist of: crop varieties, fertilizers, chemicals, fuels, labor, machinery, electricity and water. Indicator categories for farm and farmer characteristics include location, farm size, family composition, knowledge level, etc. Socio-economic indicators that have not generally been considered in the articles reviewed but have important effects on output are government policies and international markets, access to technology, employment capacity, etc. Other indicators mainly used to model sustainability include the following energy, environmental, economic and social indicators: EUE, EP, NE, energy intensiveness, GHG emissions, GWP, benefit-to-cost ratio, gross return, productivity, net return, unemployment rate, poverty, health and safety.

Jones, Antle, Basso, Boote, Conant, Foster, Godfray, Herrero, Howitt, Janssen, Keating, Munoz-Carpena, Porter, Rosenzweig and Wheeler [122] demonstrated that a minimum set of components is needed to develop agricultural-system models that are common across various applications. These include crop models that combine weather, soil, genetic and management components to simulate yield, resource use and outputs of nutrients and chemicals to surrounding water, air and ecological systems. These crop models need to account for weed, pest and disease pressures and predict performance resulting from a range of inputs and practices that represent subsistence to highly controlled, intensive production practices.

The second phase of the analysis cycle is redesigning the system. Redesign of farming systems must: (1) prepare a diversity of solutions for different futures and leave the choices to farmers and other stakeholders and (2) support farmers and other stakeholders in building their own systems, adapting them to their own situation, and deciding on their own compromises as they rely on their knowledge and experiences along with scientific knowledge [123].

Two methods for redesigning agricultural systems have been proposed: rule-based design and innovative design. Most attention has been paid to the latter. A step-by-step approach organized in a progressive transition towards more innovative systems is best [123]. System redesign begins with diagnosis followed by a determination of what steps in the system are challenging and what technical or managerial measures are needed to address them. Accordingly, evolutionary measures are designed and implemented. Then, new diagnoses are performed, and the redesign continues in the same way. These actions benefit from progress made in recent years in agricultural systemic analysis methods in situ, which make it easier to carry out precise and reliable diagnoses.

The third phase of the agricultural-systems analysis cycle is to evaluate the performance of the new system. In general, the new system is evaluated in terms of the same energy, economic, environmental and social indicators. After ensuring the desired effects of the redesign measures, the whole system is analyzed again. This cycle will continuously improve the system and move towards a more sustainable agricultural system.

4.1. Summary of Evidence

Energy Indicators Status

Our analysis revealed that tomato is the most demanding energy consumer per hectare among the crops of Iran, and sugarcane, cucumber and alfalfa follow. By contrast, sunflower is the least energy-demanding crop in Iran. The overall average TEI in farm crops of Iran during the study period was 48,029 MJ ha⁻¹. Sugar beet has the highest energy-use efficiency due to its output energy equivalent of 16.80 MJ kg⁻¹ [124]. Some crops have a negative NE, including cucumber, garlic, saffron, and tomato. The lowest EUE belongs to saffron, due to traditional production methods in Iran, in which most operations other than land preparation and fertilizer spraying are performed by human labor [52]. Further, saffron yields only about 3.7 kg ha⁻¹ under the best conditions. Although the cultivation of saffron is not efficient from the viewpoint of energy balance, it is significant economically. Khanali, Movahedi, Yousefi, Jahangiri and Khoshnevisan [52] stated that to create a better balance between the energy inputs and saffron yield, efforts should focus on increasing saffron yield and subsequently its energy productivity.

The heterogeneity in energy and environmental emissions indicators for the same product in different regions and different years is not easy to explain. Identifying reasons for these differences requires access to more information about the characteristics of agricultural systems and ecosystems. Since the studies performed in different products have usually been in different climates, the average values calculated in this article for the eight indicators can only be used in general.

4.2. Limitations

Calculating energy indicators using local rather than national or international energy equivalents would be more precise. There is a critical need to estimate local energy equivalents for inputs in each country, including Iran. Thus, energy conversion factors for food, not just agricultural products, should be developed and used at a wide range of geographic scales, from local to regional to national levels [125].

Systems analysis requires data collection that reflects a specific system's actual behavior. However, due to the vastness and complexity of agricultural systems on the ground and their geographical dispersion, collecting data from agricultural systems is challenging and costly. To overcome this difficulty, some researchers have identified innovative data collection approaches, including crowd-sourcing and remote- and close-sensing [126]. Such

innovative approaches are particularly relevant and promising for collecting data on the variability in crop management among different farms and farmers.

5. Conclusions

In this review, we examined all published papers on crop energy dynamics in Iran over a decade (2008–2018). Although many papers were published on the topic, there had been no systematic review that comprehensively collected and analyzed all results to suggest directions for future research. Energy production and consumption is costly and leads to negative environmental impacts; thus, researchers have investigated economic and environmental aspects alongside energy. In this paper, we summarized the shares of different energies across different products and regions and used DEA as an approach to find opportunities to improve the energy consumption patterns. Comparison of results of reviewed papers showed that the tomato is the most demanding energy consumer per hectare among the crops of Iran, and sugarcane, cucumber and alfalfa follow. By contrast, sunflower is the least energy-demanding crop in Iran. Our review revealed that in some published papers, values for energy and environmental indicators are reported but no explanations are given for the causes or reasons for the status in these indicators. Closer examinations of the characteristics of cropping systems should help researchers to identify and better understand energy consumption patterns. Differences in the indices for energy consumption and environmental emissions for similar products in different regions and years are not easily explained and will therefore require access to more precise information about the characteristics of agricultural systems. The meta-analysis section identifies patterns in the results of the studies included, sources of disagreement among those results, and other interesting relationships. Reasons for observed heterogeneity among the results of similar studies were interpreted using the nonparametric Kruskal–Wallis test and meta-regression. We created eight linear regression models with GHG, SE, PTE, TE, NE, TEI, EP and EUE. For meta-analysis, the data were analyzed using three factors: crop (product), year and regional climate. We can conclude that the regions have different NE, TEI and EUE. The crops also differ in terms of NE, TEI, EUE, as well as EP. No significant relationship was found between the year variable and any of the energy and environmental indicators. This result showed that over the 10-year period, cropping systems were stable in terms of energy consumption patterns and environmental emissions.

The overarching general lessons from our comprehensive review are as follows: (1) cropping systems in Iran are very diverse; (2) cropping systems are affected by a wide variety of drivers; (3) modeling and analysis of cropping systems should give comprehensive attention to factors affecting the system, and it is essential to understand the reasons for the current situation in order to devise reasonable, practical strategies for improvement; (4) cropping systems can be modeled and analyzed at different levels from field to global scales and farmers must be intimately involved in diagnosis of the problems and in devising improvement strategies; (5) analyzing cropping systems at multiple scales is recommended; (6) indicators of cropping systems can be classified into four groups, which are energy, environmental, economic and social; (7) all four groups of indicators should be used to analyze and understand cropping systems; (8) systems analysis should be done cyclically and in a step-by-step process to design corrective actions and solve problems; (9) agricultural technologies and policies (and support systems) are complementary means of improving agricultural productivity and sustainability

Author Contributions: Conceptualization, M.Z.; methodology, M.Z.; validation, S.C.; analysis, M.Z. and N.B.; investigation, N.B.; resources, N.B.; data curation, N.B.; writing—original draft preparation, M.Z.; writing—review and editing, M.Z. and S.C.; visualization, M.Z. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Informed Consent Statement: Not applicable.

Conflicts of Interest: The authors declare no conflict of interest.

Abbreviations

IE	Input Energy
EUE	Energy Use Efficiency
NE	Net Energy
FYM	Farm Yard Manure
IDE	In-Direct Energy
NRE	Non-Renewable Energy
ANOVA	Analysis of Variance
TE	Technical Efficiency
SE	Scale Efficiency
GHG	Greenhouse Gas
DE	Direct Energy
RE	Renewable Energy
TEI	Total Energy Input
PTE	Pure Technical Efficiency
EP	Energy Productivity
MPP	Marginal Physical Productivity
DEA	Data Envelopment Analysis
GLM	General Linear Model
CSA	Climate Smart Agriculture
LCA	Life Cycle Assessment
TES	Total Energy Savings
PRISMA	Preferred Reporting Items for Systematic Reviews and Meta-Analyses

References

1. Banaeian, N.; Zangeneh, M.; Clark, S. Trends and Future Directions in Crop Energy Analyses: A Focus on Iran. *Sustainability* **2020**, *12*, 10002. [\[CrossRef\]](#)
2. Mobtaker, H.G.; Keyhani, A.; Mohammadi, A.; Rafiee, S.; Akram, A. Sensitivity analysis of energy inputs for barley production in Hamedan Province of Iran. *Agric. Ecosyst. Environ.* **2010**, *137*, 367–372. [\[CrossRef\]](#)
3. Zahedi, M.; Mondani, F.; Eshghizadeh, H. Analyzing the energy balances of double-cropped cereals in an arid region. *Energy Rep.* **2015**, *1*, 43–49. [\[CrossRef\]](#)
4. Banaeian, N.; Omid, M.; Ahmadi, H. Greenhouse strawberry production in Iran, efficient or inefficient in energy. *Energy Effic.* **2012**, *5*, 201–209. [\[CrossRef\]](#)
5. Banaeian, N.; Namdari, M. Effect of ownership on energy use efficiency in watermelon farms—A data envelopment analysis approach. *Int. J. Renew. Energy Res.* **2011**, *1*, 75–82.
6. Shamsheband, S.; Khoshnevisan, B.; Yousefi, M.; Bolandnazar, E.; Anuar, N.B.; Abdul Wahab, A.W.; Khan, S.U.R. A multi-objective evolutionary algorithm for energy management of agricultural systems—A case study in Iran. *Renew. Sustain. Energy Rev.* **2015**, *44*, 457–465. [\[CrossRef\]](#)
7. Moradi, R.; Moghaddam, P.R.; Mansoori, H. Energy use and economical analysis of seedy watermelon production for different irrigation systems in Iran. *Energy Rep.* **2015**, *1*, 36–42. [\[CrossRef\]](#)
8. Khoshnevisan, B.; Bolandnazar, E.; Shamsheband, S.; Shariati, H.M.; Anuar, N.B.; Mat Kiah, M.L. Decreasing environmental impacts of cropping systems using life cycle assessment (LCA) and multi-objective genetic algorithm. *J. Clean. Prod.* **2015**, *86*, 67–77. [\[CrossRef\]](#)
9. Khoshnevisan, B.; Bolandnazar, E.; Barak, S.; Shamsheband, S.; Maghsoudlou, H.; Altameem, T.A.; Gani, A. A clustering model based on an evolutionary algorithm for better energy use in crop production. *Stoch. Environ. Res. Risk Assess.* **2015**, *29*, 1921–1935. [\[CrossRef\]](#)
10. Nabavi-Pelesaraei, A.; Abdi, R.; Rafiee, S.; Bagheri, I. Determination of efficient and inefficient units for watermelon production—a case study: Guilan province of Iran. *J. Saudi Soc. Agric. Sci.* **2016**, *15*, 162–170. [\[CrossRef\]](#)
11. Nabavi-Pelesaraei, A.; Abdi, R.; Rafiee, S. Neural network modeling of energy use and greenhouse gas emissions of watermelon production systems. *J. Saudi Soc. Agric. Sci.* **2016**, *15*, 38–47. [\[CrossRef\]](#)
12. Mohammadi-Barsari, A.; Firouzi, S.; Aminpanah, H. Energy-use pattern and carbon footprint of rain-fed watermelon production in Iran. *Inf. Process. Agric.* **2016**, *3*, 69–75. [\[CrossRef\]](#)
13. Mohammadi, A.; Tabatabaefar, A.; Shahin, S.; Rafiee, S.; Keyhani, A. Energy use and economical analysis of potato production in Iran a case study: Ardabil province. *Energy Convers. Manag.* **2008**, *49*, 3566–3570. [\[CrossRef\]](#)
14. Delshiri, A.; AghaAlikhani, M. Input-output and economic analysis of soybean production in the main cultivation areas in Iran. *Afr. J. Agric. Res.* **2012**, *7*, 4894–4899.
15. Houshyar, E.; Kiani, S.; Davoodi, M.J.S. Energy consumption efficiency for corn production utilizing data envelopment analysis (DEA) and analytical hierarchy process (AHP) techniques. *Res. Crop.* **2012**, *13*, 754–759.

16. Lorzadeh, S.H.; Mahdavidamghani, A.; Enayatgholizadeh, M.R.; Yousefi, M. Energy Input-Output Analysis for Maize Production Systems in Shooshtar, Iran. *Adv. Environ. Biol.* **2011**, *5*, 3641–3644.
17. Samavatean, N.; Rafiee, S.; Mobli, H.; Mohammadi, A. An analysis of energy use and relation between energy inputs and yield, costs and income of garlic production in Iran. *Renew. Energy* **2011**, *36*, 1808–1813. [\[CrossRef\]](#)
18. Sahabi, H.; Feizi, H.; Amirmoradi, S. Which crop production system is more efficient in energy use: Wheat or barley? *Environ. Dev. Sustain.* **2013**, *15*, 711–721. [\[CrossRef\]](#)
19. Kazemi, H.; Shahbyki, M.; Baghbani, S. Energy analysis for faba bean production: A case study in Golestan province, Iran. *Sustain. Prod. Consum.* **2015**, *3*, 15–20. [\[CrossRef\]](#)
20. Sahabi, H.; Feizi, H.; Karbasi, A. Is saffron more energy and economic efficient than wheat in crop rotation systems in northeast Iran? *Sustain. Prod. Consum.* **2016**, *5*, 29–35. [\[CrossRef\]](#)
21. Avval, S.H.M.; Rafiee, S.; Jafari, A.; Mohammadi, A. Improving energy productivity of sunflower production using data envelopment analysis (DEA) approach. *J. Sci. Food Agric.* **2011**, *91*, 1885–1892. [\[CrossRef\]](#) [\[PubMed\]](#)
22. Pishgar-Komleh, S.H.; Omid, M.; Heidari, M.D. On the study of energy use and GHG (greenhouse gas) emissions in greenhouse cucumber production in Yazd province. *Energy* **2013**, *59*, 63–71. [\[CrossRef\]](#)
23. Shah Syed Hamad, H.; Lei, S.; Ali, M.; Doronin, D.; Hussain Syed, T. Prosumption: Bibliometric analysis using HistCite and VOSviewer. *Kybernetes* **2019**, *49*, 1020–1045. [\[CrossRef\]](#)
24. Montazeri, M. Application of Multivariate Statistical Methods in Climatic Regionalization Case Study: Isfahan Province. *Geogr. Res.* **2013**, *28*, 1–16.
25. Hosseinzadeh-Bandbafha, H.; Nabavi-Pelesaraei, A.; Khanali, M.; Ghahderijani, M.; Chau, K.-W. Application of data envelopment analysis approach for optimization of energy use and reduction of greenhouse gas emission in peanut production of Iran. *J. Clean. Prod.* **2018**, *172*, 1327–1335. [\[CrossRef\]](#)
26. Pishgar-Komleh, S.H.; Ghahderijani, M.; Sefeedpari, P. Energy consumption and CO₂ emissions analysis of potato production based on different farm size levels in Iran. *J. Clean. Prod.* **2012**, *33*, 183–191. [\[CrossRef\]](#)
27. Khoshnevisan, B.; Rajaeifar, M.A.; Clark, S.; Shamahirband, S.; Anuar, N.B.; Mohd Shuib, N.L.; Gani, A. Evaluation of traditional and consolidated rice farms in Guilan Province, Iran, using life cycle assessment and fuzzy modeling. *Sci. Total Environ.* **2014**, *481*, 242–251. [\[CrossRef\]](#)
28. Mousavi-Avval, S.H.; Rafiee, S.; Sharifi, M.; Hosseinpour, S.; Notarnicola, B.; Tassielli, G.; Renzulli, P.A.; Khanali, M. Use of LCA indicators to assess Iranian rapeseed production systems with different residue management practices. *Ecol. Indic.* **2017**, *80*, 31–39. [\[CrossRef\]](#)
29. Pishgar-Komleh, S.H.; Akram, A.; Keyhani, A.; Raei, M.; Elshout, P.M.F.; Huijbregts, M.A.J.; van Zelm, R. Variability in the carbon footprint of open-field tomato production in Iran—A case study of Alborz and East-Azerbaijan provinces. *J. Clean. Prod.* **2017**, *142*, 1510–1517. [\[CrossRef\]](#)
30. Houshyar, E.; Grundmann, P. Environmental impacts of energy use in wheat tillage systems: A comparative life cycle assessment (LCA) study in Iran. *Energy* **2017**, *122*, 11–24. [\[CrossRef\]](#)
31. Mondani, F.; Aleagha, S.; Khoramivafa, M.; Ghobadi, R. Evaluation of greenhouse gases emission based on energy consumption in wheat Agroecosystem. *Energy Rep.* **2017**, *3*, 37–45. [\[CrossRef\]](#)
32. Pishgar-Komleh, S.H.; Sefeedpari, P.; Ghahderijani, M. Exploring energy consumption and CO₂ emission of cotton production in Iran. *J. Renew. Sustain. Energy* **2012**, *4*, 33115. [\[CrossRef\]](#)
33. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Mousazadeh, H. Applying data envelopment analysis approach to improve energy efficiency and reduce GHG (greenhouse gas) emission of wheat production. *Energy* **2013**, *58*, 588–593. [\[CrossRef\]](#)
34. Mobtaker, H.G.; Taki, M.; Salehi, M.; Shahamat, E.Z. Application of nonparametric method to improve energy productivity and CO₂ emission for barley production in Iran. *Agric. Eng. Int. Cigr. J.* **2013**, *15*, 84–93.
35. Asgharipour, M.R.; Mousavinik, S.M.; Enayat, F.F. Evaluation of energy input and greenhouse gases emissions from alfalfa production in the Sistan region, Iran. *Energy Rep.* **2016**, *2*, 135–140. [\[CrossRef\]](#)
36. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Mousazadeh, H.; Rajaeifar, M.A. Application of artificial neural networks for prediction of output energy and GHG emissions in potato production in Iran. *Agric. Syst.* **2014**, *123*, 120–127. [\[CrossRef\]](#)
37. Yousefi, M.; Damghani, A.M.; Khoramivafa, M. Energy consumption, greenhouse gas emissions and assessment of sustainability index in corn agroecosystems of Iran. *Sci. Total Environ.* **2014**, *493*, 330–335. [\[CrossRef\]](#)
38. Camargo, G.G.T.; Ryan, M.R.; Richard, T.L. Energy Use and Greenhouse Gas Emissions from Crop Production Using the Farm Energy Analysis Tool. *BioScience* **2013**, *63*, 263–273. [\[CrossRef\]](#)
39. Amanloo, A.; Mobtaker, H.G. Energy efficiency improvement in forage maize production using data envelopment analysis approach. *Afr. J. Agric. Res.* **2012**, *7*, 5571–5577.
40. Pishgar Komleh, S.H.; Keyhani, A.; Rafiee, S.; Sefeedpary, P. Energy use and economic analysis of corn silage production under three cultivated area levels in Tehran province of Iran. *Energy* **2011**, *36*, 3335–3341. [\[CrossRef\]](#)
41. Ghaderpour, O.; Rafiee, S.; Sharifi, M.; Mousavi-Avval, S.H. Quantifying the environmental impacts of alfalfa production in different farming systems. *Sustain. Energy Technol. Assess.* **2018**, *27*, 109–118. [\[CrossRef\]](#)
42. Mohammadi, A.; Rafiee, S.; Jafari, A.; Keyhani, A.; Mousavi-Avval, S.H.; Nonhebel, S. Energy use efficiency and greenhouse gas emissions of farming systems in north Iran. *Renew. Sustain. Energy Rev.* **2014**, *30*, 724–733. [\[CrossRef\]](#)

43. Kazemi, H.; Bourkheili, S.H.; Kamkar, B.; Soltani, A.; Gharanjic, K.; Nazari, N.M. Estimation of greenhouse gas (GHG) emission and energy use efficiency (EUE) analysis in rainfed canola production (case study: Golestan province, Iran). *Energy* **2016**, *116*, 694–700. [\[CrossRef\]](#)
44. Elhami, B.; Akram, A.; Khanali, M. Optimization of energy consumption and environmental impacts of chickpea production using data envelopment analysis (DEA) and multi objective genetic algorithm (MOGA) approaches. *Inf. Process. Agric.* **2016**, *3*, 190–205. [\[CrossRef\]](#)
45. Elhami, B.; Khanali, M.; Akram, A. Combined application of Artificial Neural Networks and life cycle assessment in lentil farming in Iran. *Inf. Process. Agric.* **2017**, *4*, 18–32. [\[CrossRef\]](#)
46. Nikkhah, A.; Emadi, B.; Firouzi, S. Greenhouse gas emissions footprint of agricultural production in Guilan province of Iran. *Sustain. Energy Technol. Assess.* **2015**, *12*, 10–14. [\[CrossRef\]](#)
47. Nikkhah, A.; Khojastehpour, M.; Emadi, B.; Taheri-Rad, A.; Khorramdel, S. Environmental impacts of peanut production system using life cycle assessment methodology. *J. Clean. Prod.* **2015**, *92*, 84–90. [\[CrossRef\]](#)
48. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Mousazadeh, H.; Sefeedpari, P. Prognostication of environmental indices in potato production using artificial neural networks. *J. Clean. Prod.* **2013**, *52*, 402–409. [\[CrossRef\]](#)
49. Nabavi-Pelesaraei, A.; Abdi, R.; Rafiee, S.; Taromi, K. Applying data envelopment analysis approach to improve energy efficiency and reduce greenhouse gas emission of rice production. *Eng. Agric. Environ. Food* **2014**, *7*, 155–162. [\[CrossRef\]](#)
50. Mohammadi, A.; Rafiee, S.; Jafari, A.; Keyhani, A.; Dalgaard, T.; Knudsen, M.T.; Nguyen, T.L.T.; Borek, R.; Hermansen, J.E. Joint Life Cycle Assessment and Data Envelopment Analysis for the benchmarking of environmental impacts in rice paddy production. *J. Clean. Prod.* **2015**, *106*, 521–532. [\[CrossRef\]](#)
51. Nabavi-Pelesaraei, A.; Rafiee, S.; Mohtasebi, S.S.; Hosseinzadeh-Bandbafha, H.; Chau, K.-W. Energy consumption enhancement and environmental life cycle assessment in paddy production using optimization techniques. *J. Clean. Prod.* **2017**, *162*, 571–586. [\[CrossRef\]](#)
52. Khanali, M.; Movahedi, M.; Yousefi, M.; Jahangiri, S.; Khoshnevisan, B. Investigating energy balance and carbon footprint in saffron cultivation—A case study in Iran. *J. Clean. Prod.* **2016**, *115*, 162–171. [\[CrossRef\]](#)
53. Mohammadi, A.; Rafiee, S.; Jafari, A.; Dalgaard, T.; Knudsen, M.T.; Keyhani, A.; Mousavi-Avval, S.H.; Hermansen, J.E. Potential greenhouse gas emission reductions in soybean farming: A combined use of Life Cycle Assessment and Data Envelopment Analysis. *J. Clean. Prod.* **2013**, *54*, 89–100. [\[CrossRef\]](#)
54. Alimagham, S.M.; Soltani, A.; Zeinali, E.; Kazemi, H. Energy flow analysis and estimation of greenhouse gases (GHG) emissions in different scenarios of soybean production (Case study: Gorgan region, Iran). *J. Clean. Prod.* **2017**, *149*, 621–628. [\[CrossRef\]](#)
55. Yousefi, M.; Khoramivafa, M.; Mondani, F. Integrated evaluation of energy use, greenhouse gas emissions and global warming potential for sugar beet (*Beta vulgaris*) agroecosystems in Iran. *Atmos. Environ.* **2014**, *92*, 501–505. [\[CrossRef\]](#)
56. Sefeedpari, P.; Shokoohi, Z.; Behzadifar, Y. Energy use and carbon dioxide emission analysis in sugarcane farms: A survey on Haft-Tappeh Sugarcane Agro-Industrial Company in Iran. *J. Clean. Prod.* **2014**, *83*, 212–219. [\[CrossRef\]](#)
57. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Yousefi, M.; Movahedi, M. Modeling of energy consumption and GHG (greenhouse gas) emissions in wheat production in Esfahan province of Iran using artificial neural networks. *Energy* **2013**, *52*, 333–338. [\[CrossRef\]](#)
58. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Yousefi, M. Prediction of environmental indices of Iran wheat production using artificial neural networks. *Int. J. Energy Environ.* **2013**, *4*, 339–348.
59. Sefeedpari, P.; Ghahderijani, M.; Pishgar-Komleh, S.H. Assessment the effect of wheat farm sizes on energy consumption and CO₂ emission. *J. Renew. Sustain. Energy* **2013**, *5*, 23131. [\[CrossRef\]](#)
60. Sheikhdavoodi, M.J.; Taki, M.; Monjezi, N. Application of artificial neural networks ANNs to predict energy output for wheat production in Iran. *Afr. J. Agric. Res.* **2013**, *8*, 2099–2105.
61. Soltani, A.; Rajabi, M.H.; Zeinali, E.; Soltani, E. Energy inputs and greenhouse gases emissions in wheat production in Gorgan, Iran. *Energy* **2013**, *50*, 54–61. [\[CrossRef\]](#)
62. Nabavi-Pelesaraei, A.; Hosseinzadeh-Bandbafha, H.; Qasemi-Kordkheili, P.; Kouchaki-Penchah, H.; Riahi-Dorcheh, F. Applying optimization techniques to improve of energy efficiency and GHG (greenhouse gas) emissions of wheat production. *Energy* **2016**, *103*, 672–678. [\[CrossRef\]](#)
63. Heidari, M.D.; Mobli, H.; Omid, M.; Rafiee, S.; Jamali Marbini, V.; Elshout, P.M.F.; van Zelm, R.; Huijbregts, M.A.J. Spatial and technological variability in the carbon footprint of durum wheat production in Iran. *Int. J. Life Cycle Assess.* **2017**, *22*, 1893–1900. [\[CrossRef\]](#)
64. Tahmasebi, M.; Feike, T.; Soltani, A.; Ramroudi, M.; Ha, N. Trade-off between productivity and environmental sustainability in irrigated vs. rainfed wheat production in Iran. *J. Clean. Prod.* **2018**, *174*, 367–379. [\[CrossRef\]](#)
65. Mohseni, P.; Borghei, A.M.; Khanali, M. Coupled life cycle assessment and data envelopment analysis for mitigation of environmental impacts and enhancement of energy efficiency in grape production. *J. Clean. Prod.* **2018**, *197*, 937–947. [\[CrossRef\]](#)
66. Nikkhah, A.; Firouzi, S.; El Haj Assad, M.; Ghnimi, S. Application of analytic hierarchy process to develop a weighting scheme for life cycle assessment of agricultural production. *Sci. Total Environ.* **2019**, *665*, 538–545. [\[CrossRef\]](#)
67. Raheli, H.; Rezaei, R.M.; Jadidi, M.R.; Mobtaker, H.G. A two-stage DEA model to evaluate sustainability and energy efficiency of tomato production. *Inf. Process. Agric.* **2017**, *4*, 342–350. [\[CrossRef\]](#)

68. Nabavi-Pelesaraei, A.; Rafiee, S.; Mohtasebi, S.S.; Hosseinzadeh-Bandbafha, H.; Chau, K.-W. Assessment of optimized pattern in milling factories of rice production based on energy, environmental and economic objectives. *Energy* **2019**, *169*, 1259–1273. [\[CrossRef\]](#)
69. Taheri Garavand, A.; Asakereh, A.; Haghani, K. Energy Consumption Flow and Econometric Models of Sugarcane Production in Khuzestan Province of Iran. *Am. Eurasian J. Agric. Environ. Sci.* **2010**, *7*, 648–651.
70. Ghorbani, R.; Mondani, F.; Amirmoradi, S.; Feizi, H.; Khorramdel, S.; Teimouri, M.; Sanjani, S.; Anvarkhah, S.; Aghel, H. A case study of energy use and economical analysis of irrigated and dryland wheat production systems. *Appl. Energy* **2011**, *88*, 283–288. [\[CrossRef\]](#)
71. Shahamat, E.z.; Asoodar, M.A.; Marzban, A.; Abdeslahi, A. Energy use and economical analysis of Sugarcane production in Iran a case study: Debel Khazaei agro-industry. *Int. J. Agric. Crop Sci.* **2013**, *5*, 249–252.
72. Taghinezhad, J.; Alimardani, R.; Jafari, A. Energy Consumption Flow and Econometric Models of Sugarcane Production in Khuzestan Province of Iran. *Atmos. Environ.* **2014**, *16*, 277–285. [\[CrossRef\]](#)
73. Mousavi-Avval, S.H.; Rafiee, S.; Sharifi, M.; Hosseinpour, S.; Notarnicola, B.; Tassielli, G.; Renzulli, P.A. Application of multi-objective genetic algorithms for optimization of energy, economics and environmental life cycle assessment in oilseed production. *J. Clean. Prod.* **2017**, *140*, 804–815. [\[CrossRef\]](#)
74. Arora, N.K. Impact of climate change on agriculture production and its sustainable solutions. *Environ. Sustain.* **2019**, *2*, 95–96. [\[CrossRef\]](#)
75. Alboghdady, M.; El-Hendawy Salah, E. Economic impacts of climate change and variability on agricultural production in the Middle East and North Africa region. *Int. J. Clim. Chang. Strateg. Manag.* **2016**, *8*, 463–472. [\[CrossRef\]](#)
76. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Mousazadeh, H. Development of an intelligent system based on ANFIS for predicting wheat grain yield on the basis of energy inputs. *Inf. Process. Agric.* **2014**, *1*, 14–22. [\[CrossRef\]](#)
77. Khoshnevisan, B.; Rafiee, S.; Omid, M.; Mousazadeh, H.; Shamshirband, S.; Hamid, S.H.A. Developing a fuzzy clustering model for better energy use in farm management systems. *Renew. Sustain. Energy Rev.* **2015**, *48*, 27–34. [\[CrossRef\]](#)
78. Mobtaker, H.G.; Akram, A.; Keyhani, A. Investigation of energy consumption of perennial alfalfa production: Case study of Hamedan province. *J. Food Agric. Environ.* **2010**, *8*, 379–381.
79. Mohammadi, A.; Rafiee, S.; Mohtasebi, S.S.; Rafiee, H. Energy inputs—Yield relationship and cost analysis of kiwifruit production in Iran. *Renew. Energy* **2010**, *35*, 1071–1075. [\[CrossRef\]](#)
80. Rafiee, S.; Mousavi Avval, S.H.; Mohammadi, A. Modeling and sensitivity analysis of energy inputs for apple production in Iran. *Energy* **2010**, *35*, 3301–3306. [\[CrossRef\]](#)
81. Mousavi-Avval, S.H.; Rafiee, S.; Jafari, A.; Mohammadi, A. Energy flow modeling and sensitivity analysis of inputs for canola production in Iran. *J. Clean. Prod.* **2011**, *19*, 1464–1470. [\[CrossRef\]](#)
82. Yousefi, M.; Mohammadi, A. Economical analysis and energy use efficiency in alfalfa production systems in Iran. *Afr. J. Agric. Res.* **2011**, *6*, 2332–2336.
83. Ghasemi Mobtaker, H.; Akram, A.; Keyhani, A. Energy use and sensitivity analysis of energy inputs for alfalfa production in Iran. *Energy Sustain. Dev.* **2012**, *16*, 84–89. [\[CrossRef\]](#)
84. Salami, P.; Keyhani, A.; Rafiee, S. The impact of farm size on energy use and profitability of red bean production in Iran: A case study in Kurdistan province. *Nat. Sci.* **2009**, *7*, 95–104.
85. Mohammadi, A.; Omid, M. Economical analysis and relation between energy inputs and yield of greenhouse cucumber production in Iran. *Appl. Energy* **2010**, *87*, 191–196. [\[CrossRef\]](#)
86. Taki, M.; Ajabshirchi, Y.; Mahmoudi, A. Application of parametric and non-parametric method to analyzing of energy consumption for cucumber production in Iran. *Mod. Appl. Sci.* **2012**, *6*, 75–87. [\[CrossRef\]](#)
87. Zangeneh, M.; Omid, M.; Akram, A. A comparative study on energy use and cost analysis of potato production under different farming technologies in Hamadan province of Iran. *Energy* **2010**, *35*, 2927–2933. [\[CrossRef\]](#)
88. Pahlavan, R.; Rafiee, S.; Omid, M. Assessing the technical efficiency in potato production in Iran. *Int. J. Green Energy* **2012**, *9*, 229–242. [\[CrossRef\]](#)
89. Moradi, M.; Azarpour, E.; Ziaeidoust, H. Study on Energy Indices for Native and Breed Rice Varieties Production in Iran. *World Appl. Sci. J.* **2011**, *13*, 137–141.
90. Pishgar Komleh, S.H.; Rafiee, S.; Sefeedpari, P. Energy use pattern and sensitivity analysis of rice production: A case study of Guilan province of Iran. *Afr. Crop Sci. J.* **2011**, *19*, 117–124. [\[CrossRef\]](#)
91. Taheri-Rad, A.; Khojastehpour, M.; Rohani, A.; Khorramdel, S.; Nikkhah, A. Energy flow modeling and predicting the yield of Iranian paddy cultivars using artificial neural networks. *Energy* **2017**, *135*, 405–412. [\[CrossRef\]](#)
92. Mousavi-Avval, S.H.; Rafiee, S.; Jafari, A.; Mohammadi, A. Optimization of energy consumption for soybean production using Data Envelopment Analysis (DEA) approach. *Appl. Energy* **2011**, *88*, 3765–3772. [\[CrossRef\]](#)
93. Mousavi-Avval, S.H.; Rafiee, S.; Jafari, A.; Mohammadi, A. The Functional Relationship Between Energy Inputs and Yield Value of Soybean Production in Iran. *Int. J. Green Energy* **2011**, *8*, 398–410. [\[CrossRef\]](#)
94. Yousefi, M.; Damghani, A.M.; Khoshbakht, K.; Veisi, H. Renewable and Nonrenewable Energy Use Pattern of Rainfed Wheat Agroecosystems in Iran. *World Appl. Sci. J.* **2011**, *13*, 1398–1403.
95. Han, H.; Wu, S. Structural Change and Its Impact on the Energy Intensity of Agricultural Sector in China. *Sustainability* **2018**, *10*, 4591. [\[CrossRef\]](#)

96. Antle, J.M.; Ray, S. Sustainability of Agricultural Systems. In *Sustainable Agricultural Development: An Economic Perspective*; Antle, J.M., Ray, S., Eds.; Springer International Publishing: Cham, Switzerland, 2020; pp. 43–94.
97. Azarpour, E.; Moraditochaee, M.; Bozorgi, H.R. Evaluation energy balance and energy indices of peanut production in north of Iran. *Afr. J. Agric. Res.* **2012**, *7*, 2569–2574.
98. Bahrami, H.; Taki, M.; monjezi, N. Optimization of energy consumption for wheat production in iran using data envelopment analysis (DEA) technique. *Afr. J. Agric. Res.* **2011**, *6*, 5978–5986. [[CrossRef](#)]
99. Tabatabaeefar, A.; Emamzadeh, H.; Varnamkhasti, M.G.; Rahimizadeh, R.; Karimi, M. Comparison of energy of tillage systems in wheat production. *Energy* **2009**, *34*, 41–45. [[CrossRef](#)]
100. Asakereh, A.; Shiekhdavoodi, M.J.; Safaieenejad, M. Energy Consumption Pattern of Organic and Conventional Lentil in Iran A Case Study: Kuhdasht County. *Asian J. Agric. Sci.* **2010**, *2*, 111–116.
101. SH, S.H.M.; Rafiee, S.; Jafari, A. A comparative study on energy and water use indicators for soybean production under different irrigation systems in Iran. *Int. J. Environ. Sci.* **2010**, *2*, 1–10.
102. Eskandari Cherati, F.; Bahrami, H.; Asakereh, A. Energy survey of mechanized and traditional rice production system in Mazandaran Province of Iran. *Afr. J. Agric. Res.* **2011**, *6*, 2565–2570.
103. Ghasemi Mobtaker, H.; Akram, A.; Keyhani, A.; Mohammadi, A. Energy consumption in alfalfa production: A comparison between two irrigation systems in Iran. *Afr. J. Agric. Res.* **2011**, *5*, 47–51.
104. Pishgar-Komleh, S.H.; Keyhani, A.; Mostofi-Sarkari, M.R.; Jafari, A. Energy and economic analysis of different seed corn harvesting systems in Iran. *Energy* **2012**, *43*, 469–476. [[CrossRef](#)]
105. Pishgar-Komleh, S.H.; Sefeedpari, P.; Rafiee, S. Energy and economic analysis of rice production under different farm levels in Guilan province of Iran. *Energy* **2011**, *36*, 5824–5831. [[CrossRef](#)]
106. Houshyar, E.; Azadi, H.; Almassi, M.; Davoodi, M.J.S.; Witlox, F. Sustainable and efficient energy consumption of corn production in Southwest Iran: Combination of multi-fuzzy and DEA modeling. *Energy* **2012**, *44*, 672–681. [[CrossRef](#)]
107. Mansoori, H.; Moghaddam, P.R.; Moradi, R. Energy budget and economic analysis in conventional and organic rice production systems and organic scenarios in the transition period in Iran. *Front. Energy* **2012**, *6*, 341–350. [[CrossRef](#)]
108. Monjezi, N.; Zakidizaji, H. Energy and Economic Analysis of Canola Production in Iran a Case Study: Khuzestan Province. *Res. J. Appl. Sci. Eng. Technol.* **2012**, *4*, 227–231.
109. Sefeedpari, P.; Rafiee, S.; Komleh, S.H.P.; Ghahderijani, M. A source-wise and operation-wise energy use analysis for corn silage production, a case study of Tehran province, Iran. *Int. J. Sustain. Built Environ.* **2012**, *1*, 158–166. [[CrossRef](#)]
110. AghaAlikhani, M.; Kazemi-Poshtmasari, H.; Habibzadeh, F. Energy use pattern in rice production: A case study from Mazandaran province, Iran. *Energy Convers. Manag.* **2013**, *69*, 157–162. [[CrossRef](#)]
111. Eskandari, H.; Attar, S. Energy comparison of two rice cultivation systems. *Renew. Sustain. Energy Rev.* **2015**, *42*, 666–671. [[CrossRef](#)]
112. Ito, J. Inter-regional difference of agricultural productivity in China: Distinction between biochemical and machinery technology. *China Econ. Rev.* **2010**, *21*, 394–410. [[CrossRef](#)]
113. Qiang, Z.; Lanying, H.; Jingjing, L.; Qingyan, C. North-south differences in Chinese agricultural losses due to climate-change-influenced droughts. *Theor. Appl. Climatol.* **2018**, *131*, 719. [[CrossRef](#)]
114. Fei, R.; Lin, B. Energy efficiency and production technology heterogeneity in China's agricultural sector: A meta-frontier approach. *Technol. Forecast. Soc. Chang.* **2016**, *109*, 25–34. [[CrossRef](#)]
115. Diao, P.; Zhang, Z.; Jin, Z. Dynamic and static analysis of agricultural productivity in China. *China Agric. Econ. Rev.* **2018**, *10*, 293–312. [[CrossRef](#)]
116. West, P.C.; Gibbs, H.K.; Monfreda, C.; Wagner, J.; Barford, C.C.; Carpenter, S.R.; Foley, J.A. Trading carbon for food: Global comparison of carbon stocks vs. crop yields on agricultural land. *Proc. Natl. Acad. Sci. USA* **2010**, *107*, 19645–19648. [[CrossRef](#)]
117. Cassman, K.G. Ecological intensification of cereal production systems: Yield potential, soil quality, and precision agriculture. *Proc. Natl. Acad. Sci. USA* **1999**, *96*, 5952–5959. [[CrossRef](#)] [[PubMed](#)]
118. Triplett, G.B., Jr.; Dick, W.A. No-Tillage Crop Production: A Revolution in Agriculture! *Agron. J.* **2008**, *100*, S153–S165. [[CrossRef](#)]
119. Lal, R. Enhancing Eco-efficiency in Agro-ecosystems through Soil Carbon Sequestration. *Crop Sci.* **2010**, *50*, S120–S131. [[CrossRef](#)]
120. Martin-Gorriz, B.; Maestre-Valero, J.F.; Almagro, M.; Boix-Fayos, C.; Martínez-Mena, M. Carbon emissions and economic assessment of farm operations under different tillage practices in organic rainfed almond orchards in semiarid Mediterranean conditions. *Sci. Hortic.* **2020**, *261*, 108978. [[CrossRef](#)]
121. Tongwane, M.; Mdlambuzi, T.; Moeletsi, M.; Tsubo, M.; Mliswa, V.; Grootboom, L. Greenhouse gas emissions from different crop production and management practices in South Africa. *Environ. Dev.* **2016**, *19*, 23–35. [[CrossRef](#)]
122. Jones, J.W.; Antle, J.M.; Basso, B.; Boote, K.J.; Conant, R.T.; Foster, I.; Godfray, H.C.J.; Herrero, M.; Howitt, R.E.; Janssen, S.; et al. Brief history of agricultural systems modeling. *Agric. Syst.* **2017**, *155*, 240–254. [[CrossRef](#)]
123. Meynard, J.-M.; Dedieu, B.; Bos, A.P. Re-design and co-design of farming systems. An overview of methods and practices. In *Farming Systems Research into the 21st Century: The New Dynamic*; Darnhofer, I., Gibbon, D., Dedieu, B., Eds.; Springer: Dordrecht, The Netherlands, 2012; pp. 405–429.
124. Asgharipour, M.R.; Mondani, F.; Riahinia, S. Energy use efficiency and economic analysis of sugar beet production system in Iran: A case study in Khorasan Razavi province. *Energy* **2012**, *44*, 1078–1084. [[CrossRef](#)]

-
125. FAO. *Food Energy—Methods of Analysis and Conversion Factors*; Food and Agriculture Organization of the United Nations: Rome, Italy, 2003.
 126. Beza, E.; Silva, J.V.; Kooistra, L.; Reidsma, P. Review of yield gap explaining factors and opportunities for alternative data collection approaches. *Eur. J. Agron.* **2017**, *82*, 206–222. [[CrossRef](#)]