

SFA vs. DEA for Measuring Healthcare Efficiency: A Systematic Review

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Abstract: Frontier techniques have been used to measure healthcare provider efficiency in hundreds of published studies. Although these methods have the potential to be useful to decision makers, their utility is limited by both methodological questions concerning their application. The aim of this paper is to search articles applying combined data envelopment analysis (DEA) and stochastic frontier analysis (SFA) in order to facilitate a common understanding about the adequacy of these methods, defining any differences in healthcare efficiency estimation and the reasons that are behind this. A systematic review of 21 such studies published the last decade was conducted. Only studies written in English were considered. Results are summarized in a form of meta-analysis in order to synthesize results and draw out further implications. Overall, DEA and SFA were found to yield divergent efficiency estimates due to many factors such as statistical noise, how inputs and outputs were defined, as well as data availability. Researchers, besides the combination of models to measure efficiency, lately have introduced environmental variables in their analyses, aiming at better understanding the relationship of these factors to efficiency and thus achieving a better decision making process. In any case the analysis concludes that there is a need for careful attention by stakeholders since the nature of the data and its availability influence the measurement of efficiency and thus it is necessary to model the behavior which generates the data by choosing the appropriate mathematical form.

Keywords: Efficiency, Data Envelopment Analysis (DEA), Stochastic Frontier Analysis (SFA), healthcare, systematic review.

1. INTRODUCTION

Healthcare costs in most developed economies have grown dramatically over the last few decades and it is widely believed that the inefficiency of healthcare institutions, at least in part, has contributed. In response to this belief, an extensive body of literature has addressed the empirical measurement of efficiency in healthcare institutions around the world. Hospitals, nursing homes, health maintenance organizations and district health authorities have been the subject of most of these efficiency studies to date. These studies share a common focus; namely, the growing volume of healthcare costs, the effect of these costs on public expenditure and private industry, and the impact of increased competition in the healthcare market. In recent decades healthcare expenditures have grown very significantly in most developed countries, from 6% of GDP in 1970 to around 11% of GDP in 2012. Many factors are behind this trend, including a demand for healthcare services that increases with income, and supply factors related with the impact of technological change [1]. This trend might hold in the future or even become more pronounced, due to, among other reasons, the phenomenon of population ageing [2]. The

control of healthcare expenditure or its financing are therefore priority aspects when public policies are designed. This priority has increased in recent years given the need to ensure the success of the fiscal consolidation processes in which most economies are immersed, following the surge in budget deficits and public debt during the economic crisis. Among the various economic policy options, those geared to attaining higher levels of efficiency in the provision of health services might be particularly appropriate. This is because, by definition, they would contribute to containing public spending (using fewer resources) while maintaining the same output and quality of the services.

These concerns have created tremendous pressure to measure the efficiency of healthcare providers and systems so that it can be evaluated and improved [3]. Despite widespread interest in evaluating efficiency, considerable uncertainty exists about whether the methods are sufficiently well developed to be used outside the research laboratory [4]. First, the term *efficiency* is used by different stakeholders to connote various constructs. Second, little is known about how well available efficiency metrics capture the constructs of interest [5]. Payers and purchasers have begun to use efficiency measures despite these uncertainties. Proponents of efficiency measurement seek to “learn

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on the job” and improve measurements through use. Those who are being evaluated on these metrics worry that the lack of conceptual clarity and the limited methodological assessments increase the likelihood that results from the metrics will create distortions in patterns of care seeking and service delivery, adding to distortions related to current payment systems [6].

The more commonly used techniques are data envelopment analysis (DEA) and stochastic frontier analysis (SFA) which employ quite distinct methodologies for frontier estimation and efficiency measurement, each with associated strengths and weaknesses. Therefore, *“...non-statistical approaches such as DEA have the disadvantage of assuming no statistical noise, but have the advantage of being non-parametric and requiring few assumptions about the underlying technology. SFA models on the other hand have the attraction of allowing for statistical noise, but have the disadvantage of requiring strong assumptions as to the form of the frontier”* [7]. DEA is favored where measurement error is unlikely to pose much of a threat and where the assumptions of neoclassical production theory are in question. Conversely, SFA should have the advantage in coping with severe measurement error and where simple functional forms provide a close match to the properties of the underlying production technology. Gong and Sickles [8] report findings along similar lines so that *“...as misspecification of functional form becomes more serious, DEA’s appeal (vis-à-vis SFA) becomes more compelling”* [8].

The aim of this systematic review was to identify and critically review the differences of existing applications of frontier techniques that have been used to measure healthcare efficiency. The review is restricted to comparisons between data envelopment analysis (DEA) and the most commonly employed parametric alternative, stochastic frontier analysis (SFA). The limitations of the empirical estimation of the efficiency measurement from the application of DEA and SFA are examined. Moreover, this paper is intended to create a common understanding among healthcare stakeholders about the adequacy of these tools to measure healthcare efficiency.

The rest of the paper is structured as follows: Section 2 provides an overview of DEA and SFA techniques for efficiency measurement. Section 3 presents the search procedure along with the typology introduced to classify the articles included in the review. Continuously, Section 3 presents the core results and outlines the findings of the current

systematic review. Sections 4 and 5 provide a discussion, draw some conclusions and provide key limitations of the current study, with the final section concluding the paper.

2. BACKGROUND

In many of the efficiency analysis studies, researchers have been interested in explaining differences in estimated efficiencies across firms or decision making units (DMU). For the efficiency analysis of DMUs, researchers have applied frontier methods such as DEA or SFA. Both methods measure inefficiency of a DMU as the distance between a best practice (or efficient) frontier and actual performance of the DMU. However, the two methods differ in some key theoretical aspects. DEA measures efficiency relative to a nonparametric estimate of an unobserved true frontier, conditional on observed data [9]. As a nonparametric method, DEA requires no assumptions about the specific form of the frontier or the probability density of inputs and outputs used in the production process. However, DEA assumes no errors and deviations from the efficient frontier rather they are entirely assumed to be due to inefficiency. Stochastic frontier models avoid some of the limitations of the DEA. Specifically, the stochastic methods allow the decomposition of deviations from the efficient frontier into a random error term that embodies statistical noise and a one-sided error term representing inefficiency. However, SFA requires the specification of a functional form for the frontier and assumptions about the distributions of the random error and inefficiency error terms which might be very restrictive [10].

DEA measures cost efficiency in two steps. First, given input prices and output levels, the cost-minimizing input vector for each hospital is calculated using linear programming. Next, cost efficiency is measured as the ratio of minimum cost to observed cost and takes a value between 0 and 1, where the value of 1 indicates a cost efficient hospital (for technical details of cost efficiency estimation [11]). The cost efficiency measures the factor by which the observed cost can be reduced if the hospital selects the optimal input bundle (which minimizes the cost of producing a given level of output given input prices) and operates at a technically efficient point (where output is produced using minimum quantities of inputs). Alternatively, cost efficiency can be estimated using SFA which, in a general form, specifies total cost as a function of outputs and input prices plus a composite error term [11]:

$$TC_i = \mathcal{F}(y_i + w_i) + \varepsilon_i$$

where TC_i represents the total cost of the i -th hospital, Y_i is a vector of outputs, W_i is a vector of input prices, and ε_i is a composite error term which can be calculated with the formula:

$$\varepsilon_i = v_i + u_i$$

where v_i captures random statistical noise, assuming that it is normally distributed, and u_i represents cost inefficiency for which a distribution (distributions assumed for the one-sided error term: half-normal, truncated-normal, exponential and gamma) must also be assumed. Given the distributional assumptions for the two error terms, the model is estimated by maximum likelihood [11]. In a cross sectional stochastic frontier model, the cost inefficiency is observed indirectly from the estimates of the composite error and is calculated as the expected value of inefficiency, conditional upon the composite residual. In the estimation of a stochastic frontier cost model, one must also specify a functional form for the cost equation. The most popular functional forms used in empirical research have been the translog and Cobb-Douglas cost functions. The translog function has been shown to be more flexible in the sense that it can provide a second-order differential approximation to any arbitrary function at a single point, making it the preferred functional form in empirical research. However, increased flexibility of the translog function comes at the cost of an increased number of parameters to estimate, and this may give rise to multicollinearity problems [11].

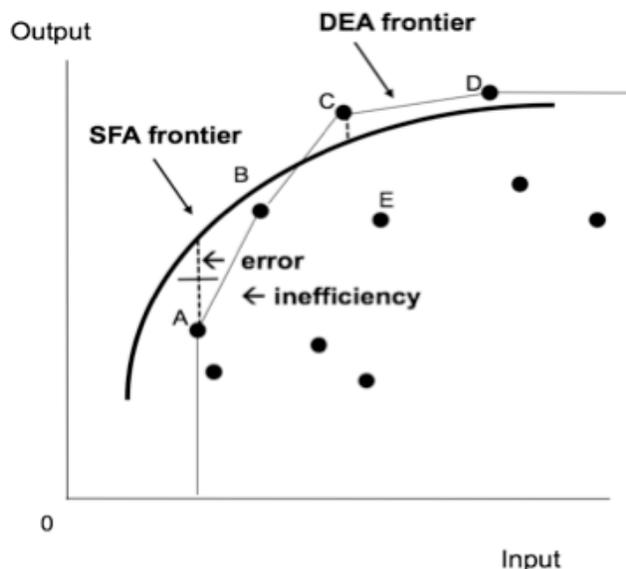


Figure 1: Illustration of SFA and DEA model.

From a policy perspective, hospital managers and policymakers can become more effective decision makers by understanding the relationships between efficiency and these two methods. The difference between DEA and SFA is described in Figure 1. Due to measurement error and other random factors affecting output, the stochastic frontier may differ from the best practice DEA frontier. For example, if the error is negative, the stochastic frontier will lie below the deterministic frontier. Using DEA we measure inefficiency as the distance from the estimated production function $f(x)$ and the x produced by x_i inputs, (measured by the angled line). Using SFA, the estimated frontier lies below and the distance from it to x is shorter (measured by the curved line). In this case, DEA will result in a higher estimate of inefficiency. Deviations from the production frontier are due to inefficiency. With SFA however, deviations may also arise from a stochastic error. The distance between the DEA frontier and the SFA frontier represents this stochastic error. If the error is positive, the stochastic frontier will lie above the DEA best practice frontier, and DEA will result in lower estimates of inefficiency (higher efficiency).

3. MATERIALS AND METHODS

3.1. Data Collection Process

During the last twenty years, non-parametric and parametric methods have been increasingly employed to measure and analyze the productive performance of healthcare services. The healthcare sector is a unique area of application, and one in which the measurement of efficiency has burgeoned over the past few years. Mortimer [7] highlighted the need for parallel application of competing methods for frontier estimation and efficiency measurement. Thus, the set of pair-wise comparisons is steadily growing as new methods for frontier estimation and efficiency measurement arise to address the shortcomings of more traditional methods. In recognition of this fact, the systematic review is restricted to comparisons between data envelopment analysis (DEA) and the most commonly employed parametric alternative stochastic frontier analysis (SFA).

The review is based on literature identified from an initial search of citation databases, review article bibliographies, and web-based resources. Following Worthington [12] and Hollingsworth [13] previous works and by adopting the steps and criteria that are summarized in Figure 2, only 21 papers, both

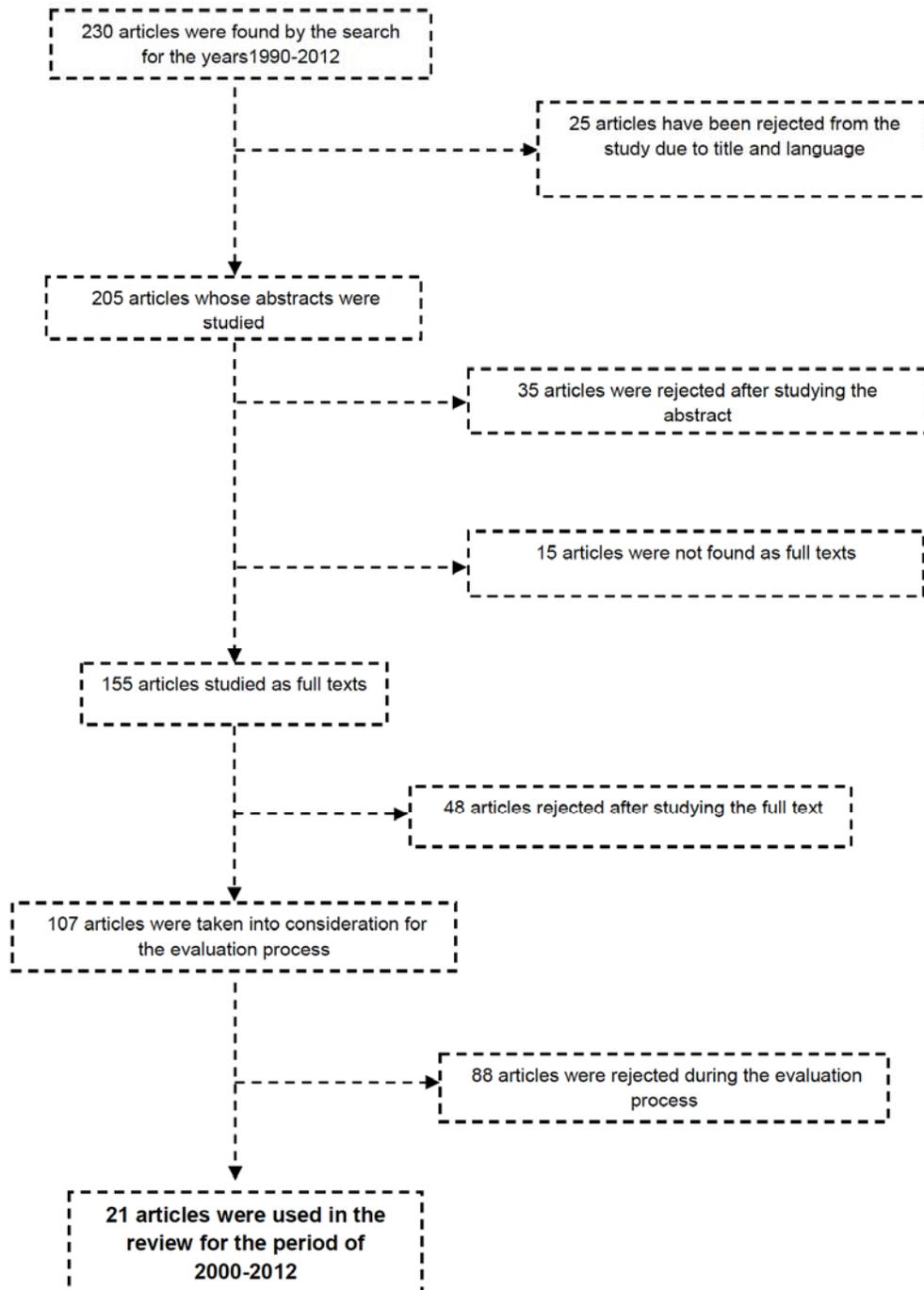


Figure 2: Flow chart of the search process.

published and working papers, were included in the review. This sample of papers, all conclude in reporting at least one pair-wise empirical comparison between DEA and SFA during the last decade 2001-2012. It should be noted that the initial search was for the period 1990-2012, but the 21 articles that were kept all refer to the last decade since the key criteria of the search was the comparison approach of DEA and SFA on efficiency measurement. The purpose is to

emphasize the fact that either there is progress in the section of controlling these techniques to measure health efficiency or these studies share the same common focus from previously analyzed.

The review that was initially started found studies for the period 1990-2012 and exclusively referred to healthcare and hospital efficiency. This paper encompasses a systematic search of all available databases, using Science Direct and Medline.

Searches were conducted using the following search terms: ‘stochastic frontier’, ‘frontier estimation’, ‘frontier techniques’. The above search terms were combined with the term ‘data envelopment analysis’ and ‘hospital efficiency’ to further focus the search, yielding 230 articles after the removal of duplicates. References were also used from these studies to identify other relevant articles.

Papers were included if they:

- described applications of both DEA and SFA for measurement hospital efficiency.
- were original or working papers.
- reported data about the reliability and/or validity of the used frontier methods.

Likewise, papers were excluded from the review:

- if they were published in a language other than English.
- if they were abstracts and not full papers.
- if they were technical reports or reviews.

By adopting these criteria and considering that the majority of the published papers that combined both DEA and SFA methods were published in the period of

2000-2012, 21 articles were identified reporting at least one pair-wise empirical comparison between both DEA and SFA. Thus, the review was focused on articles of this period by aiming to locate results of combinations of frontier techniques to evaluate healthcare efficiency.

3.2. Classification of Relevant Studies

Following the three steps of Worthington [12] and the criteria of Hussey *et al.* [14] and Mortimer [7] in their reviews, a typology to characterize the papers abstracted in the systematic review was created. This common process, as depicted in Figure 3, forms a convenient framework for the following review.

The introduced typology has four (4) levels:

- *Perspective:* what is the objective of the evaluation?
- *Efficiency measurement approach:* what mathematical approach is used?
- *Inputs and Outputs:* what inputs are used to produce the output included in the measurement?
- *Differences in efficiency results:* which is the added value of a combination of both DEA and SFA?

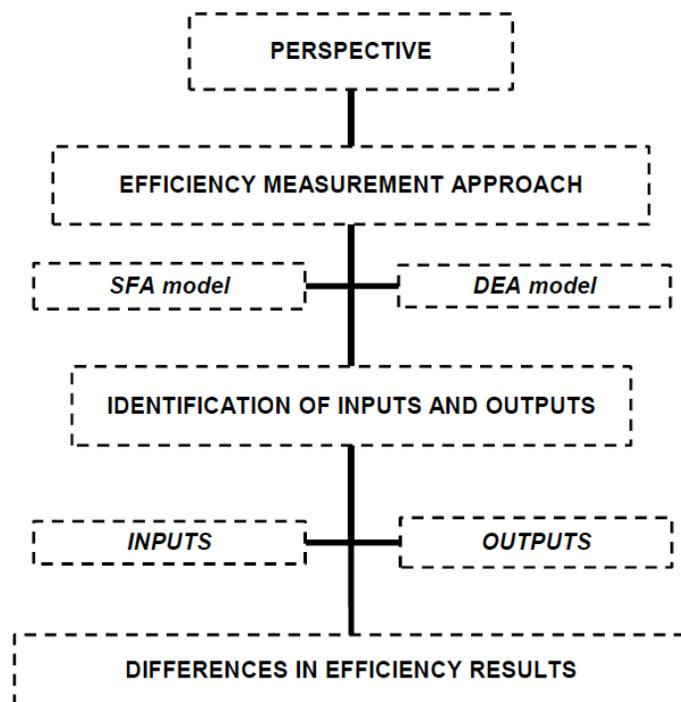


Figure 3: Typology chart.

The first tier in the typology, perspective, requires an explicit identification of the objective or rationale for the assessment. According to Hussey *et al.* [14], this element in line with the entity that is evaluating efficiency and the entity that is being evaluated, is important because different entities have different objectives for considering efficiency, have control over a particular set of resources or inputs, and may seek to deliver or purchase a different set of services.

The second tier in the typology, efficient measurement approach, involves describing the mathematical approach used. With regard to the DEA model an identification of the orientation - input or output - and Return to Scale (RTS) is required, while regarding the SFA model the mathematical form and frontier is identified. With regard to the frontier techniques, measures according to several aspects of the methodology used were characterized. For example, the type of data source, the explanatory variables included and the “scientific soundness” (reliability and/or validity) of each measure, was searched.

The third tier in the typology consists of the identification of inputs and outputs used. Outputs identify the outputs of interest and how they are measured. Two types of outputs exist: health services (e.g., visits, admissions and laboratory tests) and health outcomes (e.g., preventable deaths and clinical outcomes such as blood pressure control). On the other hand, inputs are used to produce the outputs of interest that are measured. Inputs can be measured as counts by type (e.g., number of medical staff and number of beds) or they can be monetized. The way in which inputs are measured (physical or financial) influence the way the results are interpreted. Efficiency measures that count physical inputs help to answer questions about whether the output could be produced with fewer people or fewer supplies. Efficiency measures that use financial inputs help to answer

questions about whether the output could be produced less expensively, whether the total cost of labor, supplies, and other capital could be reduced through more efficient use or substitution of less costly inputs.

Finally, the fourth tier of the typology identifies the differences of efficiency results derived from DEA and SFA analysis and the factors thought to be associated with these differences.

Thus, the papers were classified by perspective, inputs and outputs, frontier methods used, and scientific soundness.

4. RESULTS

4.1. Overview of Studies Identified

The total number of studies identified up to and including 2012 is over 230. There is a rapid increase in studies over recent years, with almost 80% of studies having been reported in the last 10 years (Figure 4). This could be explained by the fact that the economic and financial crisis has revealed the need for efficiency measurement both in the public and private sectors.

The 21 articles included in the review are presented in Table 1 in the Appendix. 42.86% of these studies are from United States followed by UK at 14.29%, while the rest of the papers are from Europe, Australia and Canada (Figure 5); the majority of the articles are simulation studies that use quite large samples of healthcare units and the range of their data cover a period of 1, 2 or 3 years with the exception of some that use 9 or 11 years of data of inputs and outputs.

Figure 6 shows the distribution of the reviewed articles towards the tier of perspective, so that over 55% of the studies are in the area of decision making and the rest are in the area of research. A significant number of studies (around 10%) investigate the extension of the used models in measuring the

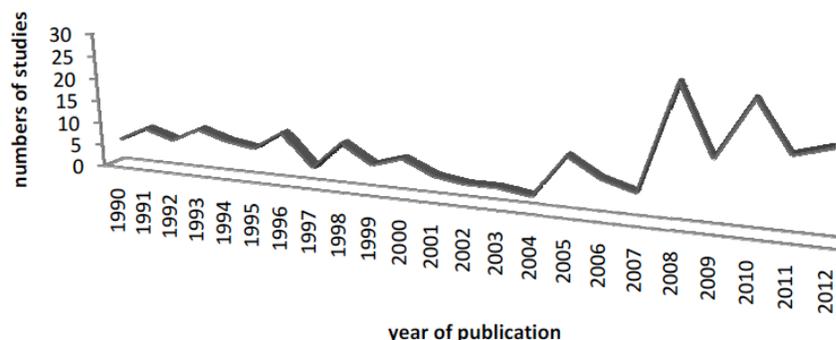


Figure 4: Number of efficiency studies 1990–2012.

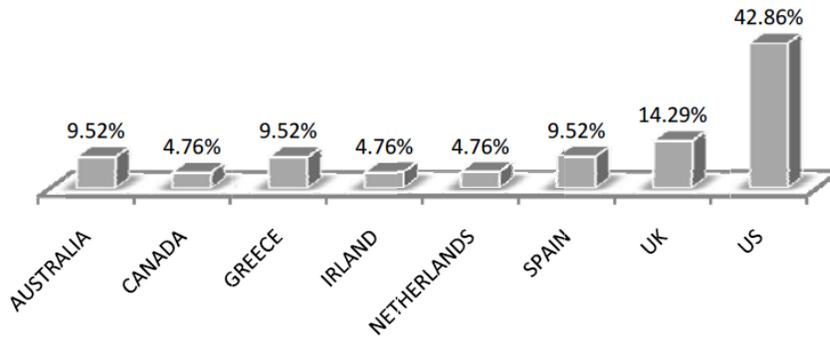


Figure 5: % Numbers of the articles by country.

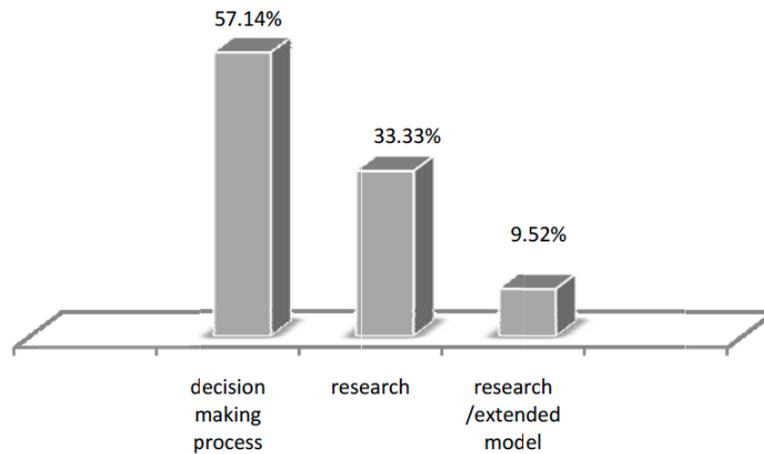


Figure 6: % of the articles by areas of perspective.

efficiency of healthcare units. The results reflect the intention of using such methods to assess the effectiveness of health units and health services in order to enhance the decision making process.

Figure 7 and 8 show that 47.62% of the studies measured health services as output variables, such as in-patient days or discharges, while inputs were measured in monetary units, indicating the objective of the researcher to answer questions regarding the less expensive production of the outputs. Continuously, 4.76% of the studies use outcome measures examining changes in health status of individuals treated, while the rest of the studies used both health services and outcome measures. Input variables are mainly measures of staff and capital employed, as examined in about 67% of the studies. Some of the studies use environmental variables for better results in the measurement of the efficiency scores. Numerous classes of these factors may influence measured levels of organizational attainment. These include: differences in the characteristics of citizens being served; the external environment – for example, geography, climate, culture; the activities of other related agencies,

both within and outside the public services; the quality of resources being used, including the capital stock;

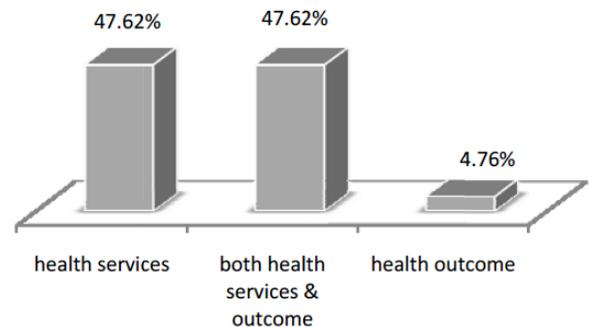


Figure 7: % of the articles included outputs.

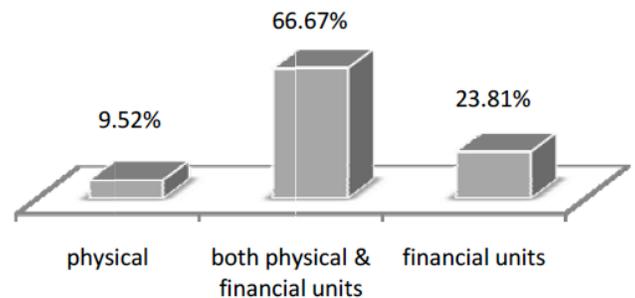


Figure 8: % of the articles included inputs.

different accounting treatments; data errors; random (or idiosyncratic) fluctuation; different organizational priorities; differences in efficiency.

With regard to the methodology used, the analysis reveals that around 45% of studies use the VRS (variable return to scale) input oriented method of DEA and the same percent use both VRS and CRS input oriented method. On the other hand, a quarter of the studies use SFA along with DEA, typically to regress factors of the efficiency scores in an attempt to determine influences on efficiency. COLS, translog analyses and Cobb-Douglas techniques are used in 25% of studies, also stochastic regression analysis and other parametric frontier techniques are used in 12% of the studies. Most of the results of the analysis highlight many divergent results between DEA and SFA methods and a small number of them reveal similar results of these techniques (Figure 9). A summary of studies comparing DEA and SFA and other parametric techniques is provided in the Appendix.

4.2. Overview of the Efficiency Measurements: DEA vs. SFA

Table 1 in the Appendix presents the 21 articles reviewed. Considering that parametric methods, such as stochastic frontier analysis (SFA), use multivariate statistical methods to explore variations in output or costs between organizations and that non-parametric method, such as data envelopment analysis (DEA), attempt to measure efficiency by estimating the optimal

level of output conditional upon the amount and mix of inputs, the articles were examined on the methodology used. The simulation studies reviewed include all possible pair-wise comparisons across the two relevant dimensions: non-parametric vs. parametric frontier estimation. In summary, the analysis reveals that both DEA and SFA have the potential to deliver biased estimates of inefficiency due to specification errors of one sort or another.

Characteristically, Giuffrida and Gravelle [15] use SFA, corrected ordinary least squares (COLS), and canonical regression as well as DEA on 90 UK FHSAs. COLS scores range from 0.868 to 0.915, stochastic frontier scores range from 0.872 to 0.982 and canonical scores range from 0.80 to 0.81, while DEA scores range from 0.904 to 0.994. They conclude that the data nature and their availability influence the measurement of the efficiency and thus it is necessary to model the behavior which generates the data by choosing the appropriate mathematical form.

Similarly, Jacobs [16] uses SFA, OLS and DEA on a sample of up to 232 UK NHS hospital trusts. The OLS mean ranges from 0.541 to 0.611, the SFA mean from 0.645 to 0.936, and the DEA mean from 0.831 to 0.876. The author concludes that the differences across methods may be due to noise and data deficiencies. Chirikos and Sear [17] using a sample of hospitals found SFA results ranging from 0.75 to 0.85, concluding that DEA and SFA yield convergent results

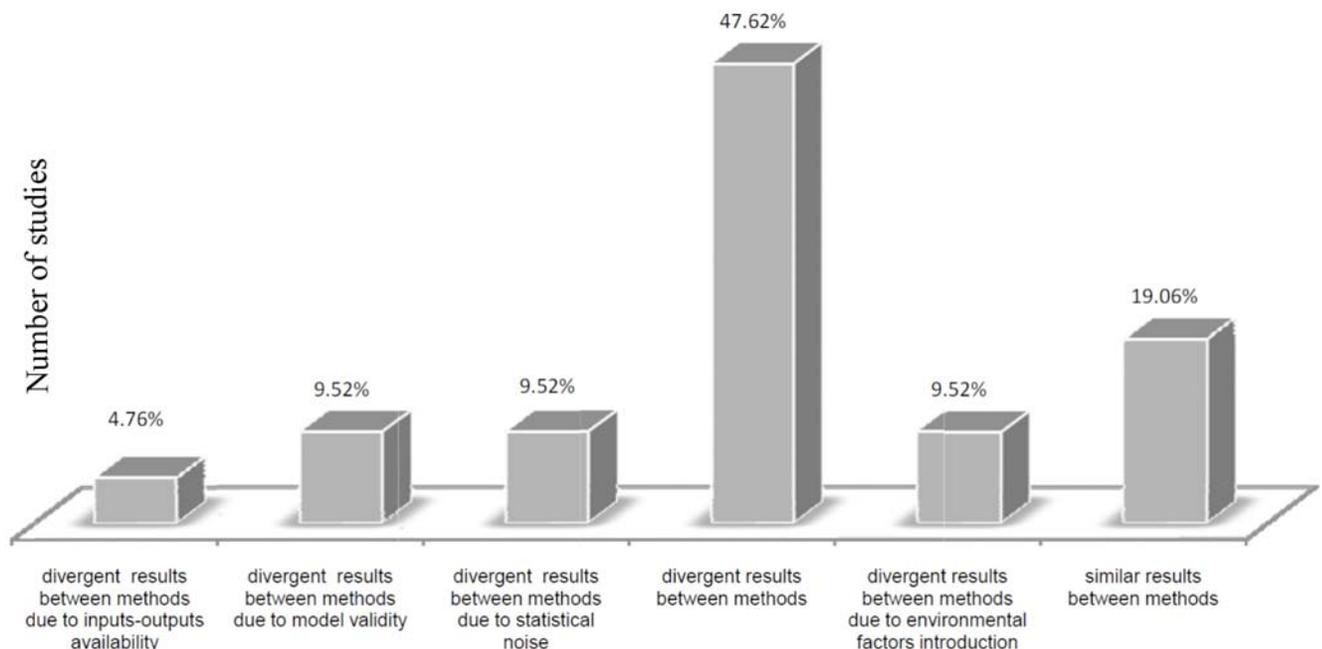


Figure 9: % of the articles by results of comparing DEA/SFA methods.

overall, but divergent results for individual hospitals, suggesting that hospital policymakers should carefully introduce these methods into the policy process. Rosenman [18] also concluded the same common views of what Chirikos [17] confirmed.

Bryce *et al.* [19] applied three models: DEA, a time varying SPF (stochastic production frontiers) and an enhanced FER (fixed-effects regression) in a sample of 585 HMOs, concluding on similar trends for the HMO industry as a whole and highlighting that these techniques are limited for either benchmarking or setting rates because the firms identified as efficient may be a consequence of model selection rather than actual performance. Likewise, Mortimer [7,20] emphasized the need for real-world comparisons to determine the relative precision and policy value of DEA and SFA. Results of his analysis showed scores ranging from 0.83 to 0.86 in DEA and 0.81 to 0.86 in SFA models, suggesting that these two techniques employ quite distinct methodologies for frontier estimation and efficiency measurement, each with associated strengths and weaknesses, such that a trade off exists in selecting the correct approach.

The last 5 years Desai [21], Smith [22], Assaf [23], Liu [24] share the same prospect that neither DEA nor SFA can be regarded as clearly dominant, and that other mixed extended methods like quantile regression, or COLS can be used and likely can yield more reliable estimates, representing useful alternatives approaches in efficiency studies. Likewise, more recent studies, Kontodimopoulos [25], Martin [26], Veen [27], Nedela [28], suggest that SFA and DEA approaches along with other techniques are viable alternatives for analyzing the impact of environmental variables and dynamic effects on hospital cost efficiency, generating similar but more consistent results in empirical application to the efficiency analysis of healthcare units. Moreover, the majority of the researchers agree on the need of being aware of using both DEA and SFA methods, by checking the robustness of the impact of environmental variables on estimated efficiency.

5. DISCUSSION AND CONCLUSIONS

This review concentrates on selected efficiency studies using frontier efficiency measurement techniques. The aim was to locate results of combinations of frontier techniques (DEA, SFA) to evaluate healthcare efficiency. Thus the review was limited to search articles applying combined DEA and SFA with the aim of defining any differences on healthcare efficiency estimation and the reasons that

are behind this. Therefore, this paper reviews empirical results drawn from published simulation studies.

The study of the 21 articles and the results, suggests that definitions of efficiency differ greatly depending on perspective, i.e., in the way that efficiency is defined as a relationship between what it costs and what service or outcome is received, rather than as a trait inherent in the provider. It is very important to address that the study seeks to focus on either the issue of extending the methods or introducing the results into the decision making process. Since the last review in this area [13, 29], the number of studies which seek to measure health service efficiency and productivity has more than doubled. The inability to measure the real output of the healthcare industry, changes in health status and the low quality of available data still leads to problems. Thus, research in this area should still be reviewed cautiously and the results of studies interpreted carefully. The techniques are still criticized, but are continually being refined. However, estimated results may still be sensitive to changes in the basic assumptions or specifications of the models used and the characteristics of the environment in which the units operate. Consequently, the results may be valid only for the units under investigation, and not necessarily be generalized.

Overall, DEA and SFA were found to yield divergent efficiency estimates due to many factors such as statistical noise and inputs and outputs definition, as well as data availability. Nevertheless, different modelling approaches have advantages and disadvantages and the choice of the most appropriate estimation method should depend on the type of organizations under investigation, the perspective taken and the quality of the available data [13, 29]. The issue of testing whether an environmental variable has a significant influence on the production process and any resulting efficiency estimates is recently overviewed. Jacobs [16], Smith and Street [30] note that the literature provides several different recommendations on how to handle such variables. From our review it is indicated that researchers, besides the combination of models to measure efficiency, introduce environmental variables in the analysis, aiming at better understanding the relationship of these factors to efficiency and thus at better decision making.

Given the limitations of frontier techniques at present it may be that they are best employed in

tandem, when possible, and if different methods suggest similar directions for results then the validity of such findings is enhanced. Since the healthcare industry is one area where efficiency measurement may have a direct policy impact, a cautious approach is necessary. As well as refining methods, the means of making efficiency results useful in a practical setting needs careful attention. Although steps are being taken in this direction there is still some way to go. The use of models with restrictions placed upon the weight given to variables, in order to reflect underlying production models or policy values, is also an interesting area requiring further research to justify the use of such restrictions. There is still room for the use of more advanced methods in efficiency measurement in the health and healthcare sector. The quality of data available for use may also be a problem to be addressed. Notwithstanding the caveats mentioned earlier regarding making comparisons across studies, and that perhaps work needs to be undertaken to think of ways of making efficiency studies comparable, these findings may have important policy implications for the organizational structure of healthcare delivery. To sum

up, careful attention should be paid to the purpose of the analysis and to how results are to be used. In particular, if they are to be used to influence economic behavior - for example in the form of setting targets, or identifying candidates for inspection - then the potential costs of making incorrect inferences should be recognized.

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APPENDIX

Table 1A: Articles Reviewed

Country	Author(s)	Sample/Period	Efficiency Techniques		Main Findings
			DEA model	SFA model	
US	Bryce et al (2000)	585 HMOs operating from 1985 through 1994	VRS input oriented	SPF (stochastic production frontier)	The estimation technique to evaluate efficient firms can affect the findings themselves.
US	Chirikos and Sear (2000)	232 acute care hospitals of Florida in continuous operation over the period 1982-1993	VRS input oriented	SFR(Stochastic frontier regression)	Hospital policymakers should not be indifferent to the choice of the frontier model used to score efficiency relationships. They may be well advised to wait until additional research clarifies reasons why DEA and SFA models yield divergent results before they introduce these methods into the policy process.
UK	Giuffrida and Gravelle (2001)	90 FHSAs (Family Health Service Authority)	CRS & VRS input oriented	COLS(corrected ordinary least squares) & stochastic frontier with half normal, exponential and truncated errors (SFN, SFE, SFT) & canonical regression (CAN).	Efficiency scores are relatively insensitive to the choice amongst regression methods or amongst DEA methods for a given year of data or a given underlying specification of the technology. The choice between DEA and regression does make a considerable difference to the rankings.
UK	Jacobs (2001)	232 UK NHS Trusts Hospitals during the years 1995/6	VRS input oriented	SCF (Stochastic Cost Frontier)	Appears to be a large amount of random 'noise' in the study which suggests that there are no truly large efficiency differences.

(Table 1A). Continued.

Country	Author(s)	Sample/Period	Efficiency Techniques		Main Findings
			DEA model	SFA model	
GREECE	Giokas (2001)	91 hospitals the year 1992	CRS & VRS input oriented	SFA (OLS estimation)	This offers practical applications for policy-makers, managers and researchers. Managers should increase in-patient days in medical care more than in-patient days in surgical care.
US	Rosenman (2001)	90 hospitals of Washington State during the years 1988 to 1993	VRS input oriented	FGLS (Feasible Generalized Least Squares procedure)	Hospitals with higher case mix indices or more beds are less efficient while for profit hospitals and those with higher proportion of medicare patient days are more efficient.
AUSTRALIA	Mortimer (2001)	38 Victoria public hospitals	VRS input oriented	SFA stochastic TC frontier Cobb-Douglas	DEA and SFA employ quite distinct methodologies for frontier estimation and efficiency measurement, each with associated strengths and weaknesses, such that a trade off exists in selecting the correct approach.
US	Ondrich and Ruggiero (2001)	simulation study of 200 & 1000 units	VRS input oriented	stochastic vs. deterministic production frontier Cobb-Douglas	Rankings for firm-specific inefficiency estimates produced by traditional stochastic frontier models do not change from the rankings of the composed errors. As a result, the performance of the deterministic models is qualitatively similar to that of the stochastic frontier models.
US	Fried et al (2002)	990 US hospital-affiliated nursing homes with data of 1993	VRS input oriented	SFR (Stochastic Frontier Regression)	The evaluation of the performance of various categories of nursing homes is dramatically altered when phenomena that are ignored in a conventional single stage DEA analysis are incorporated.
SPAIN	Ballestero (2004)	27 hospital units from Spain. Data were obtained from the regional healthcare authority for the year 1996.	CRS output oriented	SPM (single price model)	There is an ongoing issue that the validity of DEA to determine the most productive scale size appears doubtful in problems with multiple inputs and outputs. Comparison with DEA is fruitful.
IRELAND	Gannon (2004)	62 hospitals during the period 1992-1994, 61 from 1995 to 1997 and 58 from 1998 to 2000.	VRS input oriented	SFA	Comparison between DEA and SFA methods, shows that there may be high levels of random error in the data, leading to lower efficiency scores under the SFA method. Using other methods (ML) and other functional forms of the production function, the robustness of these findings to the choice of methodology for efficiency scores should be checked.
US	Desai et al (2005)	Sample of 10 units	CRS & VRS input oriented	chance-constrained formulation of DEA (stochastic model)	The simulation approach allows users to explicitly consider different data generating processes and allows for greater flexibility in implementing DEA under stochastic variations in data.

(Table 1A). Continued.

Country	Author(s)	Sample/Period	Efficiency Techniques		Main Findings
			DEA model	SFA model	
US	Smith (2005)	n/a	CRS & VRS input oriented	SFA	The assumptions underlying the derivation of weights in DEA and SFA are crucial to the judgements on efficiency they offer. At the very least, there is a suggestion that there should be a need for careful dialogue between policy makers and analysts to ensure that the methods used reflect policy requirements.
AUSTRALIA	Assaf (2008)	Sample of 101 health care foodservice operations in Australia & USA	CRS & VRS input oriented	SFA	The coefficients of the inefficiency component, estimated simultaneously with the stochastic frontier model, indicate that both the level of managers' experience and the level of managers' education are significant determinants of cost efficiency.
CANADA	Liu (2008)	sample of 100 data points	CRS & VRS output oriented	SFA model including a quantile regression	The results, based on these experiments, suggest that neither DEA nor SFA can be regarded as clearly dominant, and that quantile regression, because it yields more reliable estimates, represents a useful alternative approach in efficiency studies.
US	Lee et al (2009)	sample of 107 nursing homes from Kansas and Missouri.	VRS input oriented	SFA (COLS) corrected ordinary least squares	This analysis, which focuses on a process with much simpler measurement issues, finds evidence of inefficiency that is largely consistent with earlier studies. Increasing efficiency by adopting well-designed, reliable processes can simultaneously reduce costs and improve quality.
GREECE	Kontodimopoulos et al (2010)	sample of 124 dialysis facilities	VRS input oriented	SFA (typical Cobb-Douglas model)	Half the variation in the DEA-SFA efficiency differences was explained by environmental factors. This suggests that in addition to market instabilities, luck, and other related phenomena, decision makers in their effort to determine optimal resource allocation, should point their attention to the potentially useful insight provided by environmental factors.
UK	Martin (2010)	sample of 30 primary care trusts (PCTs) during the period 2004-06	CRS & VRS input oriented	SFA (COLS) corrected ordinary least squares	There may be substantial system-wide initiatives that could improve efficiency across the board, and this study should not lead policy makers to the conclusion that further efficiencies cannot be secured.
NETHERLANDS	Veen (2010)	data from 2 healthcare insurers (1372 observations), and 9 primary healthcare organizations (265 observations)	CRS & VRS input oriented	SFA model	The health care insurer should be cautious with drawing conclusions based on the results in this study for contracting healthcare organizations. The question that remains is what the usefulness of the efficiency assessment is, if the obtained efficiency measures are sensitive to the choice of technique.

(Table 1A). Continued.

Country	Author(s)	Sample/Period	Efficiency Techniques		Main Findings
			DEA model	SFA model	
SPAIN	Hernández (2012)	sample of 29 OECD countries between 1997 and 2009	CRS & VRS input oriented	SFA model	Empirical findings suggest that policies such as increasing the regulation of prices billed by providers and reducing the degree of gate keeping might generate efficiency gains across national healthcare systems.
US	Nedelea (2012)	760 hospitals with two years of data, 2005 and 2006	CRS & VRS input oriented	SFA model	Both methods have advantages and disadvantages that one needs to be aware of. Researchers should also consider using both methods, wherever possible, as a robustness check of the impact of environmental variables on estimated efficiency.

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