

The Relationship between Institutional Efficiency and Instructional Quality in Higher Education

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Wage setting methodologies for university faculty may be merit/market based or administered. Failure to exploit the fact that faculty productivity depends on abilities and wages results in inefficient use of university budgets. If such inefficiencies exist it suggests suboptimal productivity of the existing faculty and the inability to attract new qualified faculty. As motivation for this analysis, a simple model of university faculty “output” maximization is presented. Efficient budget allocation requires that faculty compensation be structured so that marginal productivities are equated across faculty. This paper examines and compares the efficiency of several regional universities in the US, identified as “peers”, employing the Data Envelopment Analysis (DEA) estimation method. The results suggest the existence of inefficiencies and more notably, that the homogeneity assumption regarding the peers is questionable.

Keywords: Salary; Efficiency; DEA; Higher Education; Education Administration

Introduction and Related Literature

The topic of university faculty salaries has been addressed frequently through research with varying focal points of interest. Cohn (1973) develops a multiple regression model to analyze the factors that might have an effect on faculty salaries. Although the empirical results of this study are dated and of limited applicability to the current economic situation of university professors and their pay, many of the model’s independent variables that are identified still have explanatory power today. These include the type of institution, whether it is public or private, geographic location, quality measures, institution size, and state per capita income. The variable that Cohn finds to have the greatest effect on faculty salaries is that of *institution quality*.

Tuckman and Tuckman (1976) also use regression methods “to analyze the determinants of salary structure at American universities”. The explanatory variables of their model emphasize the rate and amount that faculty publish, faculty personal characteristics such as gender and age, the university’s geographic location, and the area of faculty expertise. Two of the more definitive outcomes are that publishing and research raise salaries compared to teaching and that academic field correlates to a significant variation in faculty salaries.

Hoenack (1982) uses a theoretical approach to analyze how prices affect the choices made by faculty, students, legislators, and others involved in higher education with regard to the efficient use of resources. He concludes that inefficiency results from the existing prices, and he suggests how efficiency could be improved through *changes in prices*.

Alexander (2001) analyzes the impact that the growing disparity between faculty salaries at private universities compared to those at public universities is having on the ability of public institutions to attract and retain top-notch faculty. He describes the potential for the development of two separate higher education systems, one private and the other public. As private uni-

versities are better able to compete financially, there exists a possible “brain drain” as higher quality faculty migrate toward better salaries. The differential between faculty salaries at public and private universities is reported by Smallwood (2006) from a survey conducted by the College and University Professional Association (CUPA) for Human Resources. This survey reveals that salaries for 2005-2006 increased by 3.7 percent at private universities compared to only 3.1 percent at public universities.

In a series of articles contained in the ASHE-ERIC Higher Education Report (2001) several aspects of faculty compensation systems are addressed. The first article describes how the quality of an institution is related to how well the institution achieves its self-stated mission. Its mission is heavily dependent on the faculty, which in turn is strongly affected by the faculty compensation system. Faculty compensation is determined by a variety of factors, both external and internal. Notably, one of the internal factors cited is “market pay in the discipline”. The second article describes the three main types of faculty compensation systems: 1) the contract salary system or merit pay; 2) the single salary system; and 3) nontraditional faculty compensation systems. The third article describes the commonly used arguments in favor of either merit pay or the single salary system. The fourth article describes the advantages and disadvantages of the three compensation systems mentioned above. Interestingly, one of the primary disadvantages of the single salary compensation system is listed as “a lack of efficiency in the use of human resources”.

With regard to the methods that have been developed and utilized to measure technical and/or allocative inefficiency, the literature is replete with a history of scholarly contributions. Some of the more notable and most frequently cited of these include; Farrell (1957), Aigner et al., (1977), Kumbhakar and Wang (2006) among others.

This paper first describes a simple microeconomic model of

faculty production and budget efficiency. It then presents some preliminary efficiency results for a select group of Finance Departments at peer universities.

Theoretical Model

This model is based on the discussion in Rassouli-Currier and Currier (2008). For a given faculty member, let y denote “productivity”, i.e., number of publications, and let $x = (x_1, \dots, x_n)$ denote a vector of university expenditures on productivity enhancement. The vector x could include expenditures on wages and salaries, technology support, library facilities, conference travel support etc. The individual faculty member’s output depends on effort e and the vector x , as summarized by the “production function” $y = f(x, e)$. This production function is an increasing function of e and $x_i, i = 1, \dots, n$. In addition, the faculty member’s utility is $U(x, y) - e$ reflecting the “disutility of effort”. We assume that a (small) increase in any x_i will increase the marginal benefit of an increase in y and reduce the additional effort necessary to achieve this increase in y . Given the vector x , the faculty member selects effort (and corresponding output) to:

$$\begin{aligned} &\text{Maximize } U(x, y) - e \\ &\text{Subject to } y = f(x, e) \end{aligned}$$

This is illustrated in **Figure 1** for the case of $n = 1$. Utility increases as we move downward and to the right in the diagram. The indifference curves are concave, reflecting the assumption that a faculty member’s willingness to exert additional effort to increase productivity is highest when effort is low.

For any vector x , individual utility maximization implies solutions $e^* = e(x)$ and $y^* = y(x)$. Assuming appropriate differentiability, we have $\partial y / \partial x_i > 0, i = 1, \dots, n$. Thus, a ceteris paribus increase in any x_i will increase faculty productivity (output). Suppose now that $n = 1$, there are m individual faculty members and that the university has overall resources (e.g., salary budget) of $\$B$. Then the university budget constraint is $\sum_{j=1}^m x_j = B$. Under an administered salary program, it is essentially the case that $x_j = \frac{B}{M}, j = 1, \dots, m$. Thus, differences in rank notwithstanding, each faculty member receives the same fraction of the total budget allocation. Faculty member j ’s

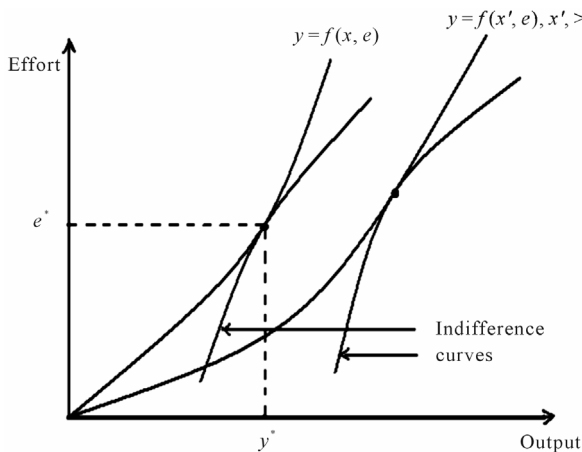


Figure 1.
Utility maximization.

output will be $y_j \left(\frac{B}{M} \right)$ with aggregate faculty output

$$\bar{Y} = \sum_{j=1}^m y_j \left(\frac{B}{M} \right).$$

Alternatively, suppose that salaries are set in such a way as to maximize total faculty output, given the university’s overall budget constraint. In this case, the university administration selects individual salaries to solve:

$$\begin{aligned} &\text{Maximize } \sum_{j=1}^m y_j(x_j) \\ &\text{Subject to } \sum_{j=1}^m x_j = B \end{aligned}$$

yielding the vector of optimal salaries $x^* = (x_1^*, \dots, x_m^*)$ and corresponding faculty output levels $y^* = (y_1^*, \dots, y_m^*)$. Since salaries are chosen to maximize total faculty output, it must be the case that $Y^* = \sum_{j=1}^m y_j^* > \sum_{j=1}^m y_j \left(\frac{B}{M} \right) = \bar{Y}$. This is illustrated in **Figure 2** where it is shown that selection of individual salary levels on the basis of the optimality criterion permits the university to operate on the “efficient” production frontier E whereas any other non-optimal salary assignment rule, measured by the number of faculty, percentage faculty holding PhD or equivalent and percentage of the tenured/on tenure-track faculty, forces the university below the efficient frontier.

The Data Set

The main source of data was the Finance Departments of 27 universities and colleges that were identified as the peers to University of Central Oklahoma in 2006 (the choice of Finance Department was arbitrary). However, the data collecting phase turned out to be extremely challenging. Universities’ published data on faculty salary and their contribution to the department is either aggregated at the college/university level or not well documented. In addition the CUPA website was not particularly helpful. Ultimately, data was collected on several variables of interest for 17 out of the 27 institutions. The list of variables and their summary statistics are presented in **Table 1**.

DEA Estimation Method

The basic idea of the DEA approach is to view universities as Decision Making Units (DMUs) with multiple inputs and outputs. DEA is a non-parametric productivity analysis model that, unlike its parametric counter-part, allows for multiple inputs and outputs to be considered simultaneously. In addition, it

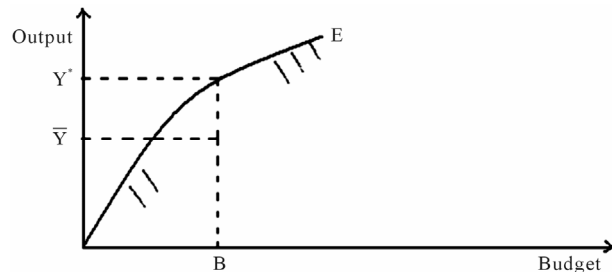


Figure 2.
Efficient production frontier.

Table 1.
Summary statistics for the variables ($n = 17$).

Variable	Mean	SD	Minimum	Maximum	Sum
<i>RA/AB</i>	42.00	40.16	2.00	134.00	714.00
<i>OPA</i>	32.41	28.83	1.00	96.00	551.00
<i>TEACH</i>	8.82	3.81	4.00	17.00	150.00
<i>PHD</i>	0.80	0.14	0.53	1.00	13.53
<i>TT</i>	0.78	0.21	0.25	1.00	13.21

Where: $Y_1 = RA/AB$ is the number of refereed articles/authored books; $Y_2 = OPA$ is other professional activities such as published articles of merit, working papers, professional presentations etc; $X_1 = TEACH$ is the number of teaching staff; $X_2 = PHD$ is the percentage of the faculty holding Ph. Ds or equivalent; $X_3 = TT$ is the percentage of the faculty that is tenured or is on tenure track.

does not require any restrictive assumptions regarding the functional form of the model.

On the down side, DEA assumes that all DMUs have the same deterministic (as opposed to stochastic) production frontier and that any deviation from the frontier is due to inefficiency, which may not be realistic. However, it is a reasonably powerful diagnostic tool that can be used to measure the efficiency of a set of homogenous DMUs individually (e.g., peer universities) relative to the most efficient unit. Studying the reasons for any possible inefficiencies and finding remedies to eliminate them is the responsibility of the unit's decision maker (Talluri, 2000).

Following Coelli et al., (1998) in the DEA method, the technical efficiency is identified as a proportional increase in the output vector with a given input vector. Therefore, the output-oriented measure of technical efficiency (in case of a production function) is the solution to the following constant returns to scale (CRS) DEA linear programming problem:

$$\begin{aligned} & \text{Maximize } \phi, \\ & \phi, \lambda \\ & \text{Subject to } -y_i + Y\lambda \geq 0 \\ & x_i - X\lambda \geq 0 \\ & \lambda \geq 0 \end{aligned}$$

where ϕ is a scalar, and y_i and x_i are column vectors of outputs and inputs respectively for the i th university. λ is an $N \times 1$ vector of constants. The variable Y is an $M \times N$ output matrix and X is a $K \times N$ input matrix, and the proportional increase in outputs that could be achieved by the i th university, holding inputs constant, is $\phi - 1$, ($1 \leq \phi < \infty$) with $1/\phi$ the university's efficiency score, which is between 0 and 1 (For a complete explanation of DEA and its advantages and disadvantages see Coelli et al., 1998)

Empirics

The empirical model for the DEA estimation is defined as Output = f (Input) where Y_1 , Y_2 are considered the outputs and X_1 , X_2 and X_3 are inputs. Due to difficulties obtaining disaggregated salary data, X_1 , X_2 and X_3 are proxies for the budget. The justification here is that the higher the faculty salary (budget), the better the universities ability to hire more faculty in general and have a larger body of teaching staff, a higher percentage of faculty holding PhDs and being tenured or on

tenure track. The DEA efficiency estimation for each institution, under the assumption of variable return to scale (VRS), was computed using DEAP 2.1 software developed by T. J. Coelli. VRS was chosen due to the rather restrictive nature of CRS. The list of the universities and their efficiency scores are not included in the paper to preserve both the privacy of the institution and because the scores in and of themselves are not the focus of this research.

The summary statistic of the efficiency scores suggests a mean score of 0.74 with a standard error equal to 0.08. The 95% confidence interval for the mean has a 0.18 margin of error. There is a wide range of efficiency from 0.023 to 1. The scores are negatively skewed (skewness = -0.99), as expected from the efficiency scores, suggesting the existence of inefficiency in at least some of the universities under study. Based on the theoretical model here, these inefficiencies stem from the existence of inefficient allocations of the total budgets.

Qualifications and Concluding Remarks

Historically, universities with larger proportions of budget allocated to teaching staff (which generally translates to a larger number of faculty) are assumed to be more productive i.e., to have higher efficiency. These universities generally have a higher percentage of faculty holding a Doctorate, tenured or on tenure track. Some examples among the sample considered here are University of Colorado-Denver and the University of Texas-San Antonio. These universities have larger number of faculty with well above average number of refereed publications. This observation does not necessarily hold for all universities.

This study attempts to get a preliminary idea of the characteristics of a "peer group" identified by UCO in 2006. The preliminary nature of this research suggests several shortcomings such as those stemming from the deterministic nature of DEA and the inability to perform statistical testing etc. However, the most important shortcoming of results stems from the enormous difficulty of the obtaining disaggregated salary data from universities. Despite all of this, the study can be considered a first step and a baseline for future studies. As one potential example, one can argue that perhaps the criteria that identify universities as peers should be reexamined since our preliminary results seem to suggest that the existence of a strong homogeneity among these universities is highly questionable. It is noteworthy that since 2007, UCO has adopted a new set of peers. This change, at least implicitly, could be a validation for the results of this paper.

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