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Identification and Analysis of Explanatory Variables for an Airline Multi-Factor Productivity Model

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Abstract

The paper aims to identify and analyze the explanatory variables for airlines productivity during 2000 2005, by testing the Pearson correlation between the single factor productivity capital, energy and labor of a sample of 45 selected international airlines (4 Brazilian carriers among them) and their productivity explanatory variables like medium stage length, aircraft load factor, hours flown and cruise speed for selected routes besides aircraft seat configuration and airlines number of employees. The research demonstrated, that a set of variables can explain differences in productivity for passenger airlines, such as: investment in personnel training processes, automation, airplane seat density, occupation of aircraft, average flight stage length, density and extension of routes, among others.

Key-words: Multifactor productivity, Multifactor productivity model, Airline productivity, Passenger airline productivity

1 Introduction

In the age of deregulation, great disparities exist between airlines in their ability to reduce unit costs by improving productivity and also to generate adequate revenues despite increasing price competition. Substantial differences exist, for example, between The United States of America's (US) airlines and non-US airlines in terms of cost efficiency, revenue generation and, in turn, profitability. Usually, measures of airline productivity to the extent they are used in the industry, are limited to relatively simple ratios - such as passenger enplanements per employee and Available Seat Miles (ASM) produced per labor dollar spent - which does not allow reliable conclusions and comparisons among the productivity of airlines.

1.1 Research objectives

This paper focuses on the productivity analysis of the main production factors for airlines: (a) labor, (b) capital and (c) energy and the identification and analysis of variables that can statistically explain these single productivities factors labor and, consequently, their Multiple Factor Productivity.

Little or no research has been done to identify variables that explain productivity of scheduled passenger airlines in order to develop a model of multiple variables, more complex to measure productivity and compare productivity between airlines.

The paper aimed to identify and discuss the explanatory variables for the productivity of scheduled international airlines by testing the Pearson correlation between the productivity changes of airlines and their explanatory variables with the objective of proposing a productivity model.

The research demonstrated that an extensive set of variables can explain differences in productivity of airlines. These variables include: investment in personnel training, process automation, airplane seat configuration, occupation of the aircraft (load factor), flight stage length, density and extension of routes, among others.

The aim of this paper was not to formulate the model itself, but to allow, from the identification of these variables, the creation of conditions to formulate such a model.

1.2 Theoretical framework

The demonstrable effects of successful US deregulation and ongoing inefficiency in the industry may have influenced the European Commission to introduce certain reforms to promote competition and thus increase the efficiency and productivity of European airlines.

Much of the literature has concentrated on productivity in the United States compared to that in Europe according to the McKinsey Global Institute (1992) and Good et al. (1993), whereas only a small proportion of papers present productivity estimates for European countries individually, as reported by Encaoua (1991). Moreover, many authors prefer to concentrate on Total Factor Productivity (TFP), as stated by Windle (1991), in favor of labor productivity measurements.

Windle (1991) compared the TFP and costs of 41 companies (14 American, 27 European and Asian), between 1970 and 1983, using the translog multilateral output index, as proposed by Caves, Christensen and Diewert (1982). In this study, five categories of inputs were utilized: (a) labor, (b) fuel, (c) flight equipment, (d) ground equipment and (e) materials. The author pointed to the evidence of a relationship between TFP and Multifactor Productivity (hereafter, MFP) input categories, such as Revenue Ton Miles (RTM) per employee.

Sickles, Good and Getachew (1994) examined the productive performance of a group of 3 East European carriers and compared them to 13 of their West European competitors during the period 1977 - 1990. The authors first modeled the multiple output/multiple input technology with a stochastic distance frontier using semi-parametric efficient methods.

Kune, Mulder and Poudevigne (2000) evaluated air transport productivity in France, Germany, United Kingdom and the United States for the period 1970-1998, using the TFP method. The objective of this study was to evaluate and compare the productivity of labor, capital and the TFP of air transport in these countries. The MFP was estimated by means of production functions and with the utilization of variables such as Value-added, Labor, and Capital.

The above mentioned authors concluded that capital is a key production factor in the airline industry and a large part of the improvement of this economic sector depends on investments in infrastructure and equipment. The differences of the capital stock per worker are also important variables for explaining performance differences between economic sectors and countries. The labor and capital productivity between France, Britain, Germany and the United States was compared in this study.

Oum and Yu (2001) produced interesting research in empirical and conceptual terms, evaluating the performance and productivity of the largest Canadian airlines for the period 1995-2000, in comparison to the eight largest American companies, using Kendrick's arithmetic index and performance metrics such as average load factor and medium stage length, evaluating also the economic and financial performance of these companies.

In Brazil, Araujo Junior (2004) studied the productivity of Brazilian airlines, during 1996 -2002, evaluating the performance of the five largest Brazilian airlines, also using Kendrick's arithmetic index and concluded that the TFP of these carriers, surpassed the average productivity of the Brazilian industry sector.

2 Methodological proposition

A Multi-factor Productivity (hereafter, MFP) index, which includes the main production factors (i.e. labor, capital and energy), was used to measure the productivity of companies surveyed during the 2000-2005 time period.

2.1 Multi-factor productivity

MFP is defined as the ratio in the quantities/volumes produced and a weighted combination of quantities and volumes of the different inputs used in the production process. Kendrick's productivity measurement method was used with changing-weight indices of outputs and inputs according to Kendrick (1996)

The MFP index is represented as the ratio between the output and input, where inputs are weighted by their share in production costs (Eq.2). The input and output values are expressed in monetary units, and therefore, MFP is a dimensionless quantity.

$$MFP = \frac{AV_t}{a_0(L_t) + b_0(K_t)} x100 \quad (1)$$

Where:

MFP = Multifactor Productivity
AV_t = Value added in a given currency;
a_0 = share of wages in the production costs;
L_t = amount of labor, in monetary terms, in period t;
b_0 = rate of return on capital invested in the same time period;
Kt = capital stock, in monetary terms, in period t.

In Eq. 1, MFP indicates the MFP index measured in monetary terms, according to Kendrick's method, which, in this case, is calculated from the ratio between the added value of the airlines in year t and the weighted relationship of labor, e.g. salaries (L_t) and capital, e.g. capital assets (K_t) in the same year, where a_0 and b_0 represent labor and capital weights, respectively.

Equation 2 is derived from Eq. 1, which makes possible to calculate productivity growth in physical terms in a time period (0, t), where AV_t is the number of passengers transported or the Revenue Seat-km (RSK); L_t represents the number of employees at the end of period t (31st December); K_t is the number of aircraft operating at the end of the same period and E_t is the amount of fuel spent also at end of period t.

$$MFP = \frac{AV_t / AV_0}{a_0(L_t / L_0) + b_0(K_t / K_0) + c_0(E_t / E_0)} x100 \quad (2)$$

Thus, Equation 2 allows us to calculate the productivity growth of a carrier in physical terms, even using different units like worked hours, number of aircraft and amount of fuel consumed since, for example, the relationship (K_t/K_0), indicates the percentage change in the aircraft fleet of a particular airline, between the time period t_1 and t_0.

Different labor and capital productivity weights, taken from Economic Report (IATA, 2001), take into account the share of input in the operational costs of carriers, according to the airline of origin, as shown in Table 1. Equation 3 gives the productivity change from a reference period 0 to a future time t.

TABLE 1: Adopted weights for Labor and Capital Productivity

Airlines	a_0	b_0
North American	0,66	0,34
European	0,72	0,28
Asian	0,57	0,43

South American	0,61	0,39

Source: IATA (2001)

Some authors, among them Moreira (1994), propose that weights a0 and b0 should be substituted, periodically, in order to reflect alterations in the production structure and changes in relative prices of capital and labor. Some organizations, such as the National Bureau of Economic Research (NBER), recommend changes every five years.

2.2 Pearson product-moment correlation coefficient

The Pearson product-moment correlation coefficient (denoted by r) is a measure of the correlation (linear dependence) between two variables X and Y, taking values from -1 through 0 to +1.

It has been used in the sciences as a measure of the strength of linear dependence between two variables. The correlation coefficient is sometimes called "Pearson's r." Pearson correlation coefficient between two variables is defined as the covariance of the two variables (X and Y) divided by the product of their standard deviations (Eq. 4):

$$\rho_{X,Y} = \frac{\operatorname{cov}(X,Y)}{\sigma_X \sigma_Y} = \frac{E[(X-\mu_X)(Y-\mu_Y)]}{\sigma_X \sigma_Y}, \tag{3}$$

Equation 3 defines the population correlation coefficient, commonly represented by the Greek letter ρ (rho). If we substitute estimates of the covariances and variances based on a sample, we obtain the sample correlation coefficient r (Eq. 5):

$$r = \frac{\sum_{i=1}^{n}(X_i-\bar{X})(Y_i-\bar{Y})}{\sqrt{\sum_{i=1}^{n}(X_i-\bar{X})^2}\sqrt{\sum_{i=1}^{n}(Y_i-\bar{Y})^2}}. \tag{4}$$

An equivalent expression gives the correlation coefficient as the mean of the products of the standard scores. Based on a sample of paired data (Xi, Yi), the sample Pearson correlation coefficient is (Eq. 5):

$$r = \frac{1}{n-1}\sum_{i=1}^{n}\left(\frac{X_i-\bar{X}}{s_X}\right)\left(\frac{Y_i-\bar{Y}}{s_Y}\right) \tag{5}$$

Cohen (1988) has observed, however, that a correlation coefficient are, in some ways, arbitrary and should not be observed too strictly.
The interpretation of a correlation coefficient depends on the context and purposes. A correlation of 0.9 may be very low if one is verifying a physical law using high-quality instruments, but may be regarded as very high in the social sciences where there may be a greater contribution from complicating factors. Pearson's correlation intervals are disclosed in Table 2.

TABLE 2 – Pearson correlation intervals

None	−0.09 to 0.0	0.0 to 0.09
Small	−0.3 to −0.1	0.1 to 0.3
Medium	−0.5 to −0.3	0.3 to 0.5
Large	−1.0 to −0.5	0.5 to 1.0

Statistical inference based on Pearson's correlation coefficient often focuses on one of the following two aims. One aim is to test the null hypothesis that the true correlation coefficient is ρ, based on the value of the sample correlation coefficient r. The other aim is to construct a confidence interval around r that has a given probability of containing ρ.

2.3 Data collection

The information and data like medium stage length, load factor, hours flown, airplane model configuration, number of employees, for the period of 2000 - 2005 was collected from international and Brazilian publications: World Air Transport Statistics (IATA), the Digest of Statistics (ICAO); Fleet and Personnel Series (ICAO), the Financial Data Series (ICAO) and the Brazilian National Civil Aviation Agency (ANAC) commercial aviation yearbook. Three categories of inputs were used: (a) labor, (b) capital, represented by flight equipment and (c) energy.

2.3.1 Labor

The labor productivity index is calculated as a multi-lateral index of 5 categories: pilots, co-pilots, other cockpit personnel, cabin attendants and other personnel. Output is composed of two separate components: scheduled revenue (passenger/km), and passengers transported.

2.3.2 Capital

Is represented by the number of aircraft used to transport passengers and cargo. In the index of aggregate capital, the percentage change in the number of aircraft was considered, adjusting it by the number of seats offered, so as to take into consideration the size of aircraft.

2.3.3 Energy

The aggregate index of energy was constructed considering the percentage change in consumption of fuel (jet fuel, since only the fleet of jets was considered).

2.4 Sampling criteria

Forty-five carriers were selected and grouped as follows:

- 26 full service
- 7 low-cost/low fare and
- 12 regional airlines

The airlines were sampled according to the following criteria:

(i) the presence and importance of the airlines in their markets (North and South American, European and Asian airlines); (ii) carriers whose data availability and previous studies indicated good operational performance and productivity were chosen.

In the case of the Brazilian airlines, a survey was undertaken through field research to collect the necessary information to:

- the four largest Brazilian airlines: TAM, GOL, VARIG, and WebJet; three of them operate in domestic and international markets, and one in the regional market and
- the Brazilian Regulatory Agency.

The single factor productivity of each of the researched companies was calculated: capital, energy and labor. These airlines single factor productivities were then compared with the explanatory variables like medium stage length, aircraft load factor, hours flown, aircraft size, aircraft seat configuration, cruising speed and aircraft engine performance for selected routes.

3 Airline productivity and explanatory variables

The purpose of this research was to understand the main variables which explain the air transport productivity, namely: labor, capital and energy productivity. These variables influence and are influenced

by others, such as investment in personnel training, processes automation, aircraft load factor, flight stage length, fleet mix, among others.

Some variables impact more than a single productivity factor. Investment in training of pilots, for instance, affects both labor and energy productivity. The flight stage length might influence both the capital and energy productivity. Airlines, however, have only limited control over some of these explanatory variables, as explained below.

3.1 Production output

Output in the airline industry is comprised of passenger services, as measured by Revenue Passenger Miles (RPMs), and cargo services, as measured by (ton x miles). Passenger miles are by far the largest component, making up more than 90% of total revenue, with the remainder attributable to ton x miles. Although the output measure does not account for changes in service quality, such as flight delays, some recent studies seem to indicate that these changes did not significantly affect output and productivity.

An airline may increase or decrease its output level through management actions, but it is usually more influenced by economic conditions, such as the demand for passengers (over which they have no control):

3.2 Average stage length

This variable depends on the route, the market structure and the air network operated by the company which, in turn, depends on the country or territory extension served, the extent of regulatory control and the attitude of government towards bilateral agreements.

3.3 Output composition or "output mix"

This variable is strongly influenced by the geographic location, the regulatory control and the different demands placed on commercial airlines. In the case of Brazilian airlines, there are small variations in the output mix. Most of them transport passengers, with a smaller share of cargo and mail.

3.4 Aircraft load factor

Some researchers, among them those of the International Labor Organization (2001), argue that the load factor is largely determined by the market demand and the extent of control the airline has over the choice of the aircraft type and the flight frequency. These researchers argue further that an airline can only manage the load factor of its fleet by adjusting the flight frequency and the aircraft size, with permission of the regulatory authorities.

3.5 Determinants of labor productivity

The labor productivity is influenced, for instance, by the amount of investment in the training of crew members (pilots and co-pilots) and maintenance teams, the outsourcing of some functions and activities, and also by the automation of some processes, such as computer ticketing.

Duke and Torres (2005) reported that, although flight crew members (which include pilots and flight attendants) are highly visible employees in the airline industry (comprising about 30% of total employment in the industry), the majority of employees work in "ground occupations." In addition to travel reservation agents and transportation ticket and customer service representatives, their occupations include aircraft mechanics, service technicians, and baggage handlers, among others.

The outsourcing of certain functions and activities, particularly those that are not concerned with the core competence of airlines, have contributed to improving labor productivity (especially so in the airline industry), as they transfer to specialized firms the rationalization of activities and processes in pursuit of a reduction in operation and service costs, such as aircraft maintenance services, ground operation support and catering.

Aircraft maintenance is currently undergoing a restructuring process. A growing number of carriers are hiring service and selling out maintenance workshops and equipment. The new technologies required for the maintenance of modern aircrafts make this activity extremely costly and a highly specialized business. Sophisticated aircraft models that require less frequent maintenance, make it increasingly difficult for an airline, individually, to justify high investments in workshops and equipment.

Currently, 75% of aircraft maintenance, according to the International Labor Organization (2001), is undertaken by airlines, while the rest is performed by specialized firms or by aircraft manufacturers. Maintenance of engines (a more specialized service), is performed, in most cases, by the manufacturers.

The IATA in its annual report (2001) forecasts for the coming years increased outsourcing for ground handling services. Currently, 75% of these services are performed by airports or airlines. In 2010 (it is estimated that) 50% of this US$ 27 billion business will be in the hands of specialists.

The IATA, in its 2001 annual report estimates that electronic ticketing ("e-ticketing") is already generating savings to airlines, every year, of about US$ 1 billion in distribution costs. The e-ticketing of airlines has now the largest sales volume on the Internet. Although the electronic sales represent a share of 5% to the conventional airlines in the United States, to some US low cost carriers, they already account for 90% of total sales.

3.6 Determinants of capital productivity

The productivity of capital is strongly influenced by the way the airlines operate their flight equipment, which, in general, represent their most important asset. The capital productivity is affected by variables such as aircraft seat density, fleet composition or mix, aircraft load factor, use of aircraft, and flight stage length.

The main fixed asset item of carriers is represented by flight equipment. The fixed assets of Brazilian airlines represented, on average, around 40% of the total assets in 2004. The main variables that impact the capital productivity of airlines are:

- average seating configuration of aircrafts, an important measure implemented by the airlines to improve productivity has been the increase of seats per aircraft. American companies (since the beginning of the 1990's), have increased the seating configuration in trunk lines.

- fleet composition or fleet mix: Brazilian carriers, similar to the American have substantially reduced the use of aircrafts that are less efficient in fuel consumption. The increase of aviation input prices (especially fuel), have forced this procedure. The fleet adequacy in terms of aircraft size, efficiency and engine output, has contributed to the increase in the capital and energy productivity, and, consequently, in the multi-factor productivity;

- fleet / aircraft operation: one of the variables with the strongest impact on the productivity of air transport is the aircraft load factor, which represents the relationship between demand and supply of passenger or freight.

Bailey, Graham and Kaplan (1985) analyzed the change of this relationship after the US market deregulation. When comparing the statistics of this study, it became evident that this relationship had changed over time, as predicted by theoretical studies.

The aircraft load factor grew with the increase in the flight stage length, according to Bailey, Graham and Kaplan (1985) based on US market data for the period between1976-1981 period, exactly the opposite of what occurred during the regulation period.

- use of aircraft: the operating objective of airlines is to use aircrafts more intensively by increasing the number of flight hours/day. The American literature based on reports of The Civil Aeronautics Board states that the equipment utilization rate increased by an average of half an hour per day in the post-deregulation period. Within Brazilian aviation (between 1995 and 2002), there was an absolute increase in the number of hours flown, despite the reduction in the number of hours flown per aircraft;

- average stage length: the flight stage is one of the operating parameters that most influence the unit cost and productivity of an airline. Airlines with flight stage beyond average have lower operating costs per unit of production.

Larger stage length means, in practice, more efficient use of flight equipment by reducing proportionally to the distance traveled, the fuel consumption, since the largest specific consumption occurs during takeoff and landing.

The Canadian experience, as reported by Oum and Yu (2001), showed that an airline with a 10% longer flight stage length had its multi-factor productivity increased in the order of 1.7%.

3.7 Determinants of energy productivity

Doganis (1985) affirms, by examining the determinants of air transport costs that the main variables that influence fuel consumption and, consequently, the energy productivity of an airline are:

- cruise speed: the cruising speed of an aircraft affects its operating cost, regardless of its size. This effect can be expressed in terms of its hourly productivity.

- aircraft size: some technological aspects have a direct effect on productivity and operating costs of each type of aircraft. Most importantly, the economic point of view is probably the size of aircraft, its cruising speed and flight range with full payload. The significance of size, speed and range of an aircraft is reinforced by the fact that these variables affect its hourly productivity, which in turn affect its operational costs.

- engine performance: the basic characteristic of an aircraft is its engine. The same type of engine may have different performances on different aircrafts and routes. The performance of an engine also depends on variables beyond the operating control of the airline: altitude and temperature of airports served, flight stage length, aircraft aerodynamics, cruising altitude etc.

4 Correlation between single-factor productivity and their explanatory variables

It is intended to identify and understand in the context of this research, the main variables that explain the multi-factor productivity and the single-factor productivity of airlines (e.g. labor, capital and energy).

We calculated the correlation matrices for each company separately and a joint matrix for the complete sample of airlines studied. Table 3 shows the correlation between the single-factor productivity of airlines and the different explanatory variables studied.

These single-factor productivities (labor, capital and energy) are dependent on other variables such as investment and training of crew members and maintenance teams, outsourcing of activities and processes, automation of administrative and operational processes, average seating configuration of operating fleet of an airline, load factor, cruise speed, stage length, among others (Fig. 1).

Some variables affect more than a productivity factor. The investment in the training of pilots for instance, influences both labor and energy productivity. The flight stage length influences both capital and energy productivity.

As it can be seen in Table 3, there is, in descending order, a large positive correlation, according to the intervals defined in Table 1, between capital productivity and cruise speed (r^2=0.9405), capital productivity growth and seat configuration/density (r^2 = 0.9062).

TABLE 3 - Correlation matrix between single-productivity factor of airlines and its main explanatory variables.

	Stage Length	Load Factor	Aircraft Use	Seat Configur.	Cruise Speed	Employees

	[km]	[%]	[hours flown]	[seats/aircraft]	[km/h]	[unit]
Capital productivity	0.3039	0.6575	0.6320	0.9062	0.9405	Very low correlation
Energy productivity	0.4774	0.2623	0.4033	-0.8229	-0.8764	Very low correlation
Labor productivity	Very low correlation	0.2397	Very low correlation	Very low correlation	Very low correlation	-0.6078

The study showed that the correlation capital productivity versus cruise speed is larger among carriers with larger average stage length and, in the case of the correlation capital productivity versus aircraft size (seats/aircraft), the larger correlation occurs among airlines with larger aircraft size.

There is also a large positive correlation between capital productivity and load factor (+0.6775), and capital productivity versus aircraft utilization in hours flown (+0.6320). In the case of the productivity of energy, a large but negative correlation between energy productivity and cruise speed (-0.8764) and energy productivity and seat density (-0.8229), and between labor productivity and number of employees (-0,6078) was verified.

The study showed also in the case of the correlation energy productivity versus cruise speed that this correlation is larger among airlines operating with lower cruise speed that can be explained by the aircraft engines consumption which increases proportionally higher with the increasing speed of aircraft.

A medium positive correlation, according to the criteria defined in Table 2, can be inferred from energy productivity growth and stage length (+0.4774) and energy productivity growth and aircraft utilization (+0.4033). And finally a small positive correlation is verified between energy productivity growth and load factor (+0.623) and labor productivity and load factor (+0.2397).

The correlation labor productivity versus number of employees of airlines was negative (-0.6078), which was expected. Regarding the correlation of productivity labor and other variables like stage length, load factor, aircraft utilization, aircraft size and cruising speed it can be considered low (correlation with load factor) and very low.

The analysis of the determinants of the single productivity factors labor, capital and energy and their explanatory variables led us to the conceptual model as shown in Fig. 1, which reproduces the inter-relationship between the main productivity elements labor, capital and energy and their explanatory variables. Figures in brackets, in Fig. 1, represent the Pearson correlation between explanatory variables and the respective single-factor productivity, as shown in Table 3.

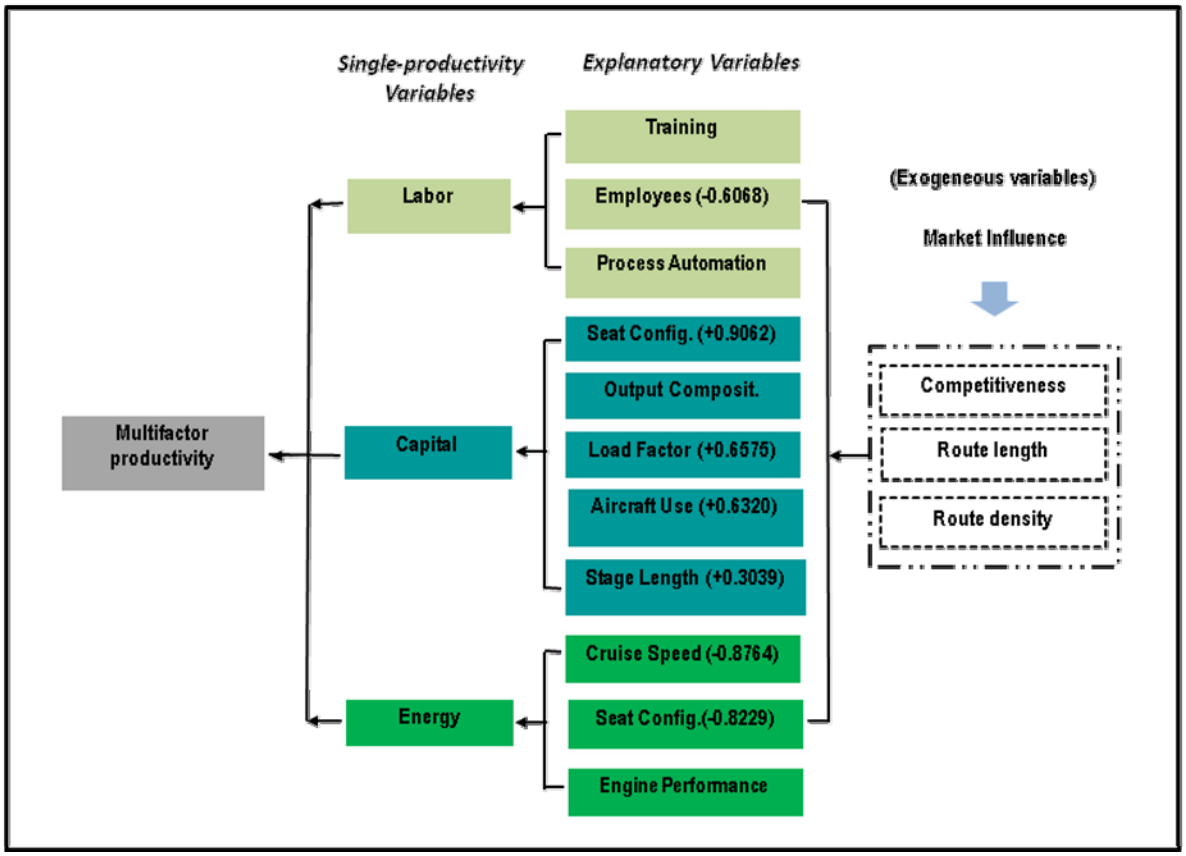

FIGURE 1 – Productivity model for scheduled airlines, deduced from the collation and analysis of labor, capital and energy productivity and their explanatory variables.

5 Conclusion

The survey was conducted with 41 international airlines within the categories Full Service Companies (FSC), Low Cost/Low Fare (LCC) and Regional Companies (RC) between 2000 and 2005 (and as part of this sampling, the four major Brazilian airlines were included).

Kendrick's productivity method was used to measure the multiple-factor productivity growth of linear dependence between the single-factor productivity of labor, capital and energy and the productivity explanatory variables of the airlines analysed.

The largest negative correlation was found between energy productivity and cruise speed (correlation of 88%), and between labor productivity and number of employees (correlation of 60%).

Among the variables which best explain the MFP are those related to the factor capital, seating configuration of the aircraft (positive correlation 90.6%) and average operating speed (positive correlation 94.0%), which also impacts negatively the energy factor, and with respect to labor, the number of employees used in the operation. The research demonstrated that capital is a key productivity factor in the airline industry and a large part of the productivity improvement on this economic sector, depends on investments in fleet, infrastructure and equipment.

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