

Analysis of absorptive capacity and foreign direct investment in the productivity of Brazilian municipalities

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Abstract

Paper aims: to measure the impact of Foreign Direct Investment (FDI) and Absorptive Capacity (AC) on productivity in Brazil.

Originality: no articles were found using the AC as moderator of FDI in Brazil.

Research method: the Malmquist Index was employed and the Feasible Generalized Least Squares (FGLS) model was chosen, which has greater robustness for correcting the heterocedasticity and autocorrelation.

Main findings: productivity is more impacted by FDI than Economic Complexity. There is a two-year time lag to the impact of the FDI on productivity. The interaction between FDI and AC becomes positive only when experience and schooling are controlled. The number of PhD employees was better as proxy for AC than R&D workers.

Implications for theory and practice: there was no positive relationship between productivity and the number of higher education professionals.

Keywords

Malmquist index. Economic complexity. Higher education. Panel Data.

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1. Introduction

Absorptive Capacity (AC) is one of the most influential concepts in management literature. First introduced by Cohen & Levinthal (1989) and then developed by Zahra & George (2002) when it comes to learning and innovation on a company, and currently is a key word for a variety of strategies, routines and learning processes that influence the company's ability to exploit the external knowledge needed to build other organizational capacities (Todorova & Durisin, 2007; Zahra & George, 2002).

The diffusion and acquisition of knowledge determine companies' potential for innovation (Griliches, 1998). However, Absorptive Capacity is necessary to understand and transform external knowledge flows, essential to achieve innovation and growth of companies (Cohen & Levinthal, 1990). It is noteworthy that the focus of many studies is about organizational factors that facilitate or inhibit the transfer process, including the organization's Absorptive Capacity (Gonzales & Martins, 2015).



Lapan & Bardhan (1973) argue that companies need a certain level of Absorptive Capacity before they can benefit from technologies developed by other companies. Cohen & Levinthal (1989) argue that increasing Research and Development (R&D) activity increases efficiency indirectly, making it faster to assimilate technologies developed elsewhere.

The ability to attract FDI, according to Sari et al. (2016), can bring great benefits to a receiving country, since incoming multinational corporations provide direct and indirect benefits to the host country's economy.

The direct benefits of foreign subsidiaries can take form as new investments, productive capacity, demand for labor, demand for intermediate goods and, sometimes, exports that stimulate national income or economic growth, providing new opportunities and increasing tributary revenue (Takii, 2005).

In organizations, knowledge is already taken for granted as a strategic asset, as well as an explanatory variable for its performance and growth (Grant, 1996). Given this context, Foreign Direct Investment (FDI) generates effects on the productivity of sectors that receive FDI directly, thus, it is believed that FDI is a determining factor for increasing productivity and efficiency.

Therefore, in an increasingly globalized and dynamic world, it is necessary to analyze the impact of FDI on receiving economies, especially the effects related to economic growth. Many papers such as Ubeda & Pérez (2017), Li-Ming et al. (2016), Kim (2015), Girma (2005) investigated whether the effects of FDI on productivity growth depends on Absorptive Capacity.

Migueluez & Moreno (2015) affirm that Absorptive Capacity is an essential element for the regions to make the most out of the knowledge and information flows that arrive, allowing them to obtain productivity gains and competitive advantages.

In addition to the effects of Foreign Direct Investment and Absorptive Capacity, several studies have demonstrated the importance of productive diversification on the countries' economies (Hidalgo & Hausmann, 2009; Hartmann, 2014). Economic Complexity is an explanatory factor of productivity because it considers that the goods produced are generated from the knowledge available in a region. Therefore, for a region to be more complex, there must be knowledge, more skilled workers and technology, factors that increase productivity and generate economic growth.

Given this context, the aim of this article is to measure the impact of Foreign Direct Investment and Absorptive Capacity on Productivity of 106 Brazilian cities between 2010 and 2014. In this context, the problem to be investigated is: *How much is the impact of the FDI on municipal level productivity in Brazil? Is there a time lag to FDI affects productivity? Is there an interaction effect between FDI and Absorptive Capacity on productivity? Does the FDI impact productivity more than the region's complexity?*

This article analyzes the causality between FDI, absorptive capacity and economic complexity over productivity. The FDI and absorptive capacity contribute to increase productivity because greater external investment helps in the emergence of new businesses and products, as absorptive capacity assists in the introduction of new practices or techniques in companies. However, we argue that there is an interaction between FDI and AC, because productivity increase would require more investment and also available knowledge. This is the phenomenon that this article seeks to analyze in order to investigate the importance of both practices on the economy.

Foreign Direct Investment (FDI) has increased significantly over the past two decades. This significant increase has received special attention in discovering the relation between the inflow of FDI and economic growth in the receiving country. Thus, it is important to find out not only the factors that impact the incoming volume of FDI in a given economy, but also the effects of such capital on economic growth of the receiving region, since these impacts may be conditioned to the host AC that may directly affect regional productivity.

Thus, the article presents an econometric model for FDI and productivity moderated by AC, as well as comparing models and AC proxies through Information Criteria. Likewise, the investigation is able to detect which is the best model that explains the productivity, and further analyzing the best proxy for AC. This paper also contributes to verify if there is an interaction between FDI and AC and to compare the impact of FDI for Economic Complexity, which directly influences policy recommendations for the problem investigated.

The present paper is organized in five sections besides this introduction. The second and third sections present the theoretical reference on FDI and its relation with productivity, respectively. In the fourth section there is the research method and in the fifth section the results and discussions. Finally, the main considerations are found in the sixth section of this paper.

2. Foreign direct investment and absorptive capacity

The relationship between the institutional environment and the decision to invest in multinationals was originally proposed by Basi (1963) based on the argument that the quality of political institutions influences FDI (Amal & Seabra, 2007).

The Foreign Direct Investment, according to Anyanwu (2017), can be a catalyst for economic diversification, helping to overcome countries in relation to excessive dependence on natural resources. In addition, the FDI is an important form of long-term capital, which aids investment in infrastructure and other initiatives for economic development.

International investments can be classified as direct investments and portfolio investments. External investments in portfolio correspond to capital flows that are not oriented towards the operational control of the company that receives external capital. On the other hand, FDI refers to an investment that seeks to acquire a lasting interest in a company whose exploration takes place in another country (International Monetary Fund, 1998). In addition, the OECD considers a foreign investment to be a direct investment when it holds a capital holding of at least 10% and may influence the management of the receiving enterprise (Organisation for Economic Co-operation and Development, 1987).

The ability to attract FDI can bring great benefits to the receiving country as incoming multinational corporations provide direct and indirect benefits to the host economy. The direct benefits of foreign affiliates can take the form of new investments, productive capacity, demand for labor, demand for intermediate goods and sometimes exports that stimulate national income or economic growth, provide new opportunities and increase tax revenue (Takii, 2005).

The determinants of the FDI are evaluated based on the microeconomic variables which are directly related to the firm's performance, in addition to the macroeconomic and institutional variables, which show the specific factors of the receiving country of foreign investments. Thus, the theory of industrial organization represents a change in the definition of determinants of FDI (Santos et al., 2015).

The main determinants of external investments in productive activity are characterized by the sources of raw materials, transport structure or costs, by the factorial endowment of countries or, ultimately, by comparative advantages as Krugman & Obstfeld (2010) report. There are also other factors that encourage the attraction of FDI, which are the employed technological level, vertical production processes, market size, institutional and macroeconomic stability.

The paper of Economou et al. (2017) examined the determinants of FDI flow in 24 OECD countries and 22 developing (non-OECD) countries using standard fixed effects panel as well as a dynamic panel approach. The first set of empirical results indicates market size, trade openness, unit labor cost, schooling, taxation, gross capital formation and institutional variables as important determinants of FDI for OECD countries. The relevant results for the sample of developing countries indicate the size of the FDI market, labor cost, gross capital formation and institutional variables as significant determinants of FDI.

Kumari & Sharma (2017) analyzed the main determinants affecting the flow of FDI in developing countries using an unbalanced panel data set (1990-2012). The findings reveal that market size, trade openness and human capital have a positive association with FDI inflows. This study has significant implications for policy makers, managers and investors. Policy makers will be able to understand the importance of the key determinants of FDI mentioned in the document and take steps to formulate policies that encourage FDI.

According to Kim (2015), the AC may act as a role to alleviate the negative effects of the FDI on a company's productivity. In fact, studies have suggested the existence of positive knowledge spillovers linked especially to technology and its impacts on other related variables, as well as negative effects of spillover (Lewison, 1991; Howells, 2005).

3. Productivity and FDI

Productivity measures the degree of efficiency with which a given economy uses its resources to produce consumer goods and services (Messa, 2013). Increasing productivity is the fastest way to achieve economic growth and social welfare, since such production gains reflect the full effectiveness of the productive sector as well as a society's degree of development (Felema et al., 2013).

Thus, a greater the Economic Complexity Index (ECI) of the countries means that the economy is more diversified, products with higher technology emerge and more skilled workers are hired. This directly affects productivity (Hidalgo & Hausmann, 2009; Hartmann, 2014). In this context, the first hypothesis follows:

H_1 : ECI positively impacts Brazil's municipal level productivity.

This hypothesis of a positive relationship between complexity and productivity in Brazil is important, since it is a way of making the economy more competitive, generating more economic growth and development for the country. If this hypothesis is confirmed, it is possible to recommend economic diversification as a means of achieving greater product growth than through natural resources.

Thus, Total Factor Productivity (TFP) is intended to indicate how efficiently the economy combines all its resources to generate output. From this conceptualization, the dynamics of the indicator is a result of local technological progress (Messa, 2013). This means getting more output with the same amount of resources employed or using fewer resources to achieve the same production level.

The concept of TFP is also applied to explain why one country develops more than the other, as the development and average income of people can be credited to the efficient use of inputs.

According to Porcile et al. (2005), TFP stands out among productivity indicators because it is a multifactor indicator. The TFP allows us to identify and dissociate the share of output change that can be attributed to efficiency gains and the share that can be attributed to the accumulation of inputs (capital, labor, and human capital).

TFP can be calculated using the Malmquist Index (MI). Several papers have employed the Malmquist Index as a proxy for TFP, such as Cao et al. (2017), Mu et al. (2017), Fu & Ji (2017) and Chen et al. (2016).

MI was initially suggested for analyzes of input use (Malmquist, 1953), and was subsequently applied in a productivity context by Caves et al. (1982) and, in addition to Farrell's (1957) ideas on efficiency measures, led to the Malmquist productivity index developed by Färe et al. (1994). The first paper written by Aigner & Chu (1968) applied linear programming methods to the analysis of panel data of the social area and then measures of growth of the TFP as the sum of the component of change of efficiency and technical change (Coelli et al., 1998).

Caves et al. (1982) emphasize MI's ability to measure temporal changes in total productivity of production factors in terms of efficiency changes and technological changes, relative to the units considered. As described by Färe et al. (2004), total productivity is defined by the changes of the factors of production as changes in the total product generated from the various inputs used, and define the technological changes as movements of the efficiency frontier in a given period.

Thus, it can be said that the advantage of using this variable as a proxy for TFP is that it evaluates the productivity indexes over time, decomposing them into indexes that capture variation of technical efficiency and technological changes.

The authors Dupasquier & Osakwe (2003), Anyanwu (2006), Abor et al. (2008) and Inekwe (2013) argued that FDI can play an important role in development of countries efforts including the transfer of modern technologies and the enhancement of local workforce skills, resulting in increased productivity.

Girma (2005) reports that negative impacts on TFP are due to the allocation of FDI in AC. Lucas (1988) shows that FDI flows contribute to economic growth in recipient countries, increasing capital stock and repercussions. The repercussions of FDI can increase the stock of knowledge in investor countries (Li & Liu, 2005). However, there is a maturation period to allow the inflow of foreign financial resources to positively impact productivity. From this evidence, one can elaborate the hypotheses:

H_2 : FDI and productivity are directly proportional for Brazilian municipalities.

H_{2a} : The FDI time lag is statistically significant and makes the model more robust to explain productivity.

The H_2 and H_{2a} hypotheses are important for future policy recommendations because if this relationship is proven for Brazil and depending on the size of the impact of complexity and the FDI on productivity, public policy makers can develop mechanisms to increase complexity or to bring foreign investment into the country.

Coming from that concept, literature has argued that productivity gains will be tied to AC and innovative classes' learning. In this way, innovative producers are more receptive to new technologies and thus are able to maximize gains and reduce costs (Felema et al., 2013). In this way, another hypothesis is corroborated:

H_3 : FDI moderated by AC has a more significant effect than FDI without being moderate.

In this sense, Hypothesis 3 can demonstrate that it is not enough to develop mechanisms to increase complexity and receive more external resources. This is because the AC is necessary to raise productivity. In this sense, in addition to the financial capital needed to generate more productivity, it would also be necessary to guarantee the existence of human capital (education, workers with doctorates, or R&D workers).

Table 1 shows some determinants of the TFP selected in the literature in order to help choosing the control variables of the model of this paper.

According to the determinants of productivity raised by the literature, some hypotheses can still be proposed according to the availability of data:

H_4 : The average age of workers has a significant impact on productivity.

H_5 : The number of workers with higher education has an impact on productivity.

Table 1. TFP determinants according to literature.

Authors/Year	Region	Period	TFP determinants
Danska-Borsiak (2018)	Czech Republic, Hungary, Slovakia and Poland	2000-2013	R&D activities; infrastructure; physical capital; structural change; financial system; location of the region; per capita income
Otsuka (2017)	Japan	1980-2010	Social capital; populational crowding
Otsuka & Natsuda (2016)	Malasya	2000-2004	FDI; R&D; human capital; employed technology
Kim (2016)	Japan	1973-2010	Exports; imports; R&D; income; work quality; work hours
Akinlo & Adejumo (2016)	Nigeria	1970-2009	Commercial opening; Foreign Direct Investment; inflation; human capital; unemployment rate
Harris & Moffat (2015)	Great Britain	1997-2008	Real gross output; real intermediate inputs; employment; capital; age; single plant
Giovanis & Ozdamar (2015)	USA	1976-2009	Age; size; short-term indebtedness; long-term indebtedness; liquidity; Index of added value; Active relationship for sales; risk proxy; market share; Business entry; exit companies; Industry average growth
Arazmuradov et al. (2014)	Ex-USSR countries	1995-2008	GDP; human capital; FDI; Import of machinery and equipment
Castiglionesi & Ornaghi (2013)	Spain	1990-2006	Index of new technologies use; income; Percentage of R&D staff in total work force; Quotas of students with higher education in relation to the total workforce; human capital; R&D expenses
Sheng & Song (2012)	China	1998-2007	Participation in R&D; market share; Herfindahl Index; Export quota
Dańska-Borsiak & Laskowska (2012)	Poland	2003-2009	Human capital level; R&D; Investments
Kim (2011)	Malasya	2002-2004	Work; capital; training cost per skilled worker; qualified worker; number of employment with higher education; R&D

In this sense, hypotheses 4 and 5 may show the need to control some characteristics of workers in order to obtain better econometric estimates.

4. Method

This section presents the research method applied in the development of the research, the study population, definition of the variables, as well as the description of the Malmquist Index and econometric models.

4.1. Data

This article used macroeconomic variables from 106 Brazilian cities (see the Annex 1), randomly selected and which had exporting companies between 2010 and 2014. This unique database differentiates this article from other studies previously developed, since few studies on this topic have been developed in the cities of Brazil and other emerging economies.

The selected cities that are representative of all Brazilian regions are described in Annex 1. It is worth noting that, of the whole sample, 48.11% (51 cities) belong to the Southeast region, 29.25% (31 cities) to the South region, 13.21% (14 cities) to the Northeast, 6.60% (7 cities) to the Central-West region, and 2.83% (3 cities) to the North.

To achieve the proposed goal, two steps are required: Calculating the Malmquist Index; and the estimation of econometric models. The data were collected from the Instituto Brasileiro de Geografia e Estatística (IBGE), Instituto de Pesquisa Econômica Aplicada (IPEA), Fundação SEADE, and Relação Anual de Informações Sociais (RAIS/MTE).

The definitions of the variables used in each step are given in Table 2.

4.2. The malmquist index

This paper had as its first step the calculation of the Malmquist Index, used as a proxy for Total Factor Productivity (TFP).

Table 2. Variables used in the two steps.

Variable	Definition
Malmquist Index	Proxy for Total Factor Productivity - Calculated by means of the variables municipal GDP (output), average income of workers and fixed assets (inputs).
Municipal GDP	Gross Domestic Product, that is, everything that was produced in a certain region during a certain period of time (Instituto Brasileiro de Geografia e e Estatística, 2018).
Average income of workers	Average monthly salary of formal workers (Instituto Brasileiro de Geografia e e Estatística, 2018).
Fixed assets	Assets necessary for the operation of the company, for example, buildings, machinery, land, etc.
ECI – Economic Complexity Index	Proxy for the Economic Complexity or productive diversification of a country or region. The greater the number of technological products produced in a region, the greater the complexity of the OEC (Observatory of Economic Complexity).
FDI – Foreign Direct Investment	Index measured by the ratio of the number of international companies in the municipality by the total number of companies in it.
Number of formal workers with doctorate	Proxy for Absorptive Capacity. Professionals with doctorates are linked to the activities of decision making, development of new routines and working methods, as well as the development of new products and markets.
Number of formal workers in R&D	Proxy for Absorptive Capacity. R&D professionals are linked to innovation departments as well as to the development of new products and markets.
Age of the worker	The average age of workers in each municipality was used as a proxy for the experience in the labor market.
Number of workers with higher education	Proxy for the number of skilled workers in each municipality.

The Malmquist Index, originally proposed by Malmquist (1953) and Caves et al. (1982), has the purpose of comparative analysis of the relative performance of a set of Decision Making Units (DMUs) - namely, the municipalities, in adjacent periods of time, when appropriating the respective data of inputs and products of a base period - is therefore the elaboration of an index number applied to the performance analysis (Färe et al., 2004).

As described by Färe et al. (2004), we define the changes of Total Factor Productivity as changes in the total product generated from the various inputs used, and define the technological changes as efficiency frontier movements in a given period of time. In this way, the modification of the first one is decomposed by the changes computed in efficiency and in technology, as described in Equation 1:

$$ATFP = TA \times EA \tag{1}$$

Given that the Malmquist Index (MI) - representing the Alteration in the Total Factor Productivity, ATFP, of a DMU, over a certain period of time - TA, Technological Alteration - measured from the change of the efficiency frontier in the same time interval - and EA, the Efficiency Alteration - measured from the change in the relative performance of a DMU, in the same time horizon considered for the other components.

In a complementary manner, Färe et al. (2004) demonstrate that it is possible to specify the calculation of Total Factor Productivity changes from an integrated perspective between the Malmquist Index and the Data Envelopment Analysis (DEA) models. The DEA, in turn, is a non-parametric approach with a mathematical programming model, which is often used to evaluate the relative efficiency of units with multiple inputs and outputs (Ferreira & Gomes, 2009).

In this way, the index can be calculated by Equation 2:

$$MI = \sqrt{\left(\frac{D^0(x^0v, y^0v)}{D^t(x^t v, y^t v)} \frac{D^0(x^0v, y^0v)}{D^t(x^0v, y^0v)}\right)} \cdot \left(\frac{D^t(x^t v, y^t v)}{D^0(x^0v, y^0v)}\right) = TA * EA \tag{2}$$

On what:

MI = Malmquist Index;

D° = Distance function relative to the period 0 border;

D^t = Distance function relative to the boundary of period t;

y°v = Quantity of DMU virtual output under analysis in period 0;

x°v = DMU virtual input quantity under analysis in period 0;

y^tv = Quantity of DMU virtual output under analysis in period t;

x^tv = Quantity of virtual input of the DMU under analysis in period t;

$D_0(x^0, y^0)$ = Distance from the DMU in period 0 relative to the border of period 0;

$D_0(x^t, y^t)$ = Distance from the DMU in period t relative to the border of period 0;

$D_t(x^0, y^0)$ = Distance from the DMU in period 0 relative to the border of period t;

$D_t(x^t, y^t)$ = Distance from the DMU in period t relative to the border of period t;

TA = Technological Alterations of a DMU between periods 0 and t; and,

EA = Efficiency Alterations of a DMU between periods 0 and t.

The results offer subsidies for the productivity analysis, since they allow to identify if there was increase in technological progress, improvement in the total efficiency of the DMU, or both. Thus, this procedure becomes appropriate to identify if the changes in the development of an environment were related to the technological changes or the total productivity of the factors of production of the DMUs, which in this case are the municipalities. One can say that technological change is a set of possible combinations that expand or contract, determined by the technology environment.

In order to calculate IM, two inputs (Average income of workers and capital assets) and one output (municipal GDP) will be used, with orientation to the output CRS (Constant Returns to Scale), since it is possible to establish proportionality between inputs and outputs, that is, to increase GDP with the increase of Income and Capital (Cao et al., 2017; Mu et al., 2017; Fu & Ji, 2017; Chen et al., 2016).

4.3. Econometric models

The multiple regression analysis, according to Gujarati & Porter (2011), concerns the prediction of the dependent variable based on two or more independent variables. In this way, the form of the multiple linear equation is an extension of simple linear regression, in which other variables (X_1, X_2, \dots) are inserted. Equation (3) of the multiple regression is defined by:

$$Y = \beta_0 + \beta_1 X_1 + \dots + \beta_n X_n + \epsilon \quad (3)$$

In which Y represents the dependent variable and $X_1 \dots X_n$ represent the explanatory variables, ϵ represents the experimental error (independent random errors), and β_0, β_1 and β_2 are the regression coefficients.

It is worth mentioning that the dependence of a variable Y (the dependent variable) on other X variables (the explanatory variable) is rarely immediate. Too often, Y responds to X with time lapses. This lapse is called time lag (Gujarati & Porter, 2011). Thus, some variables were lagged to analyze the maturation time for impact on the productivity of the municipalities.

In order to test the hypotheses elaborated in the previous section, the following panel data econometric models were formulated using log-linear regressions, since it is possible to interpret parameters as elasticities.

Model 1: ECI positively impacts Brazil's municipal level productivity (H_1) and FDI and productivity are directly proportional for Brazilian municipalities (H_2)

$$\ln y_{it}^{MI} = \beta_0 + \beta_1 \ln ECI_{it} + \beta_2 \ln FDI_{it} + \epsilon_{it} \quad (4)$$

Model 2: The FDI time lag is statistically significant and makes the model more robust to explain productivity (H_{2a})

$$\ln y_{it}^{MI} = \beta_0 + \beta_1 \ln ECI_{it} + \beta_2 \ln FDI_{it} + \beta_3 \ln FDI_{it-1} + \beta_4 \ln FDI_{it-2} + \epsilon_{it} \quad (5)$$

Model 3 and 4: FDI moderated by AC has a more significant effect than FDI without being moderate (H_3)

$$\ln y_{it}^{MI} = \beta_0 + \beta_1 \ln ECI_{it} + \beta_2 \ln FDI_{it} + \beta_3 \ln FDI * AC_{it} + \epsilon_{it} \quad (6)$$

$$\ln y_{it}^{MI} = \beta_0 + \beta_1 \ln ECI_{it} + \beta_2 \ln FDI_{it} + \beta_3 \ln FDI * AC_{it} + \epsilon_{it} \quad (7)$$

For model 3, the number of professionals with a PhD is considered to be the CA proxy, and for model 4, the number of R&D professionals in the municipality.

Model 5: The average age of workers has a significant impact on productivity (H_4) and the number of workers with higher education has an impact on productivity (H_5)

$$\ln y_{it}^{MI} = \beta_0 + \beta_1 \ln ECI_{it} + \beta_2 \ln FDI_{it} + \beta_3 \ln FDI * AC_{it} + \beta_4 \ln AGE_{it} + \beta_5 \ln SCHOOL_{-it} + \epsilon_{it} \quad (8)$$

In what: The *lnMI* is Malmquist Index; *lnECI* represents the municipality's Economic Complexity Index; *lnFDI* is the natural logarithm of Foreign Direct Investment; *lnFDI*CA* represents the Direct Foreign Investment moderated by Absorptive Capacity (as proxy for Absorptive Capacity, the number of professionals with a PhD and the number of R&D professionals available in the municipality were used: directors, managers and technicians); *lnAGE* represents the average age of workers in the municipality; *lnSCHOOL* represents the level of schooling (number of workers with higher education).

4.3.1. Panel data technique

In panel data, the same cross-sectional unit is monitored over time. In summary, panel data have a spatial and a temporal dimension (Gujarati & Porter, 2011). In this way, as this paper includes the time limit from 2010 to 2014 for several municipalities, it is therefore a balanced panel analysis.

Panel analysis can be performed through two basic models: Fixed and Random Effects models. The Fixed-Effect model removes time invariant characteristics of the explanatory variables so that the liquid effect can be analyzed, while the Random Effects model has the same assumptions as the fixed effects model, that is, the intercept varies from one individual to another, but not over time, and the response parameters are constant for all individuals and for all periods of time (Gujarati & Porter, 2011).

In the fixed-effects model, the Least Square Dummy Variable (LSDV) estimator will be consistent and efficient (Holland & Xavier, 2005). Each individual in the sample has its own characteristics that may or may not influence the explanatory variables. Thus, it is assumed that some information may provide bias or impair the explicability of the variables (Greene, 2011). On the other hand, the RE model assumes that there is no correlation between the individual effects and the other random variables, and its estimation is carried out through Generalized Least Squares (GLS) (Holland & Xavier, 2005).

According to the assumptions of the econometric models, the error term (ε_{it}) must have a constant variance and the error terms should not be correlated with each other. Accordingly, the violation of these hypotheses may induce autocorrection and heteroscedasticity (Greene, 2011).

Thus, this paper applies more advanced econometric techniques to correct these problems for panel data, since the hypothesis of time lags for the FDI can induce to severe autocorrelation. Thus, Feasible Generalized Least Squares (FGLS) method is chosen, which incorporates an AR (1) structure at the stochastic disturbance (Greene, 2011).

Another problem of econometric models is multicollinearity. Multicollinearity is defined as the presence of a high degree of correlation between the independent variables (Freund et al., 2006). To verify the presence of multicollinearity, the VIF (Variance Inflation Factors) test was calculated. The FIV can be calculated by the following equation:

$$VIF_j = \frac{1}{1-R_j^2} \quad j = 1, 2, \dots, p \quad (9)$$

Where: p is the number of predictor variables; R_j^2 is the multiple correlation coefficient, resulting from the regression of X_j in the other $p-1$ regressors.

The choice of the appropriate model from the statistical point of view is an extremely important topic in data analysis (Bozdogan, 1987). In order to compare the robustness of each model, Akaike Information Criterion (AIC) and the Bayesian Information Criterion (BIC) (Akaike, 1973, 1974; Schwarz's, 1978; Judge et al., 1985) was employed. The model with lower value of AIC and BIC is considered the best fit model.

Finally, we emphasized that all the estimates and statistics present in the aforementioned method were calculated through Matlab® and Stata® softwares.

5. Results and discussion

Through the Variance Inflation Factor (VIF) test, there was no evidence of multicollinearity. Besides that, we performed the White Test before measuring the regressions to verify the presence of heteroscedasticity (Greene, 2011). We found autocorrelation and heteroscedasticity in our estimations. For this reason, we estimate our models through the Feasible Generalized Least Squares (FGLS) procedure that incorporates an AR (1) structure at the stochastic disturbance.

Table 3 shows the estimation for Model 1, which measured the effect of FDI on productivity. OLS estimation are inconsistent, as previously discussed in the method. In addition, although the Hausman test was favorable

Table 3. Estimates for MI and FDI (Model 1).

Variables	(1) OLS	(2) FE	(3) RE	(4) FGLS
<i>LnECI</i>	0.0192* (0.0116)	0.0111 (0.0364)	0.0205 (0.0151)	0.0123** (0.00500)
<i>LnFDI</i>	0.219*** (0.0541)	-0.155 (0.157)	0.201*** (0.0531)	0.185*** (0.0264)
<i>Constant</i>	-0.675*** (0.0993)	-1.392*** (0.307)	-0.711*** (0.117)	-0.706*** (0.0529)
Observations	270	270	270	268
R-squared	0.141	0.005		
Number of num		70	70	68

Robust standard errors in parentheses. *** p<0.01; ** p<0.05; * p<0.1.

for FE estimation, the FGLS model has greater robustness by correcting heterocedasticity and autocorrelation in the model.

Model 1 measured the effect of Economic Complexity and Foreign Direct Investment on productivity. The estimates show that the Economic Complexity positively impacts (1.23%) the productivity of the Brazilian municipalities, presenting statistical significance of 5%. This result confirms the Hypothesis 1 proposed in this study and agrees with previous results found by Felema et al. (2013).

Economic Complexity is related to greater capacity of production of goods, and in a diversified environment workers are more engaged and willing to deliver than expected by people, and conflicts that can hinder productivity are less recurrent.

Another result for Model 1 is the FDI analysis, which has a positive impact (18.5%) on productivity, and a statistical significance of 1%, which confirms the Hypothesis 2: FDI and the Malmquist index are directly proportional. Note that FDI has a greater impact on productivity than the complexity of municipalities. This is an important result, since it demonstrates that Foreign Direct Investment in the Brazilian regions is an important determinant for the increase of productivity.

This result also shows that there is a greater effect of FDI on productivity than Economic Complexity. For this reason, estimates show that it is more important to develop public policies that encourage the entry of foreign capital, as a way of generating new businesses, markets and products. The indirect effect of this policy would also be to increase the Economic Complexity and its impact on productivity, in view of the productive diversification and the spillover effect in the regions.

Our results corroborate with the studies by Girma (2005), Sari et al. (2016), Liu (2008), Pessoa (2008), Baldwin et al. (2005), Takii (2005) and Javorcik (2004), which indicate that Foreign Direct Investment generates improvements in the use of technologies, management methods, new product and process development, knowledge transfer and corporate productivity.

The results of the other models are shown in Table 4. Model 2 analyzed the time lags effect for FDI in relation to productivity. Estimates show that by controlling the dissipation of the FDI effect on productivity, the FDI of year zero maintains the expected signal but loses statistical significance. In the analysis of the 1 year lag time, the effect is not statistically significant. Only after two years of the entry of Foreign Direct Investment, that is, the lag team of 2 years, there is statistical significance of 10% and positive impact (0.0788).

This result confirms Hypothesis 2a of this study, since when comparing the information criteria of Model 1 (AIC = 266.0159 and BIC = 276.8112) with model 2 (AIC = 83.0699 and BIC = 97.4839), it is verified that the Model 2 further explains the phenomenon between FDI and the Malmquist Index, showing greater robustness.

Thus, it is possible to say that there is a maturation period so that the inflow of financial resources from abroad has a significant impact on the productivity of municipalities. It is also possible to argue that two types of public policies can be developed in Brazil. First, a short-term policy, focused on raising Economic Complexity and productive diversification, which has an immediate effect on the economy. Second, a medium-term policy (two-year time lag), so that the inflow of foreign capital would generate more productivity. Combined, these policies would raise productivity by 9.33%.

According to the results of models 3 and 4, FDI moderated by Absorptive Capacity (proxies of number of professionals with doctorates and number of R&D professionals, respectively) although presented high statistical significance (1%) for both models, the obtained signal was not what expected, which refutes the Hypothesis 3

Table 4. Estimates for the FGLS models.

MI	Model 1	Model 2	Model 3	Model 4	Model 5
<i>LnECI</i>	0.0123** (0.00500)	0.0145*** (0.00119)	0.0178*** (0.00519)	0.0215*** (0.00555)	0.0106** (0.00468)
<i>LnFDI</i>	0.185*** (0.0264)	0.0552 (0.0414)	0.200*** (0.0257)	0.226*** (0.0296)	0.108*** (0.0346)
<i>LnFDI(1lag)</i>	-	0.0417 (0.0445)	-		
<i>LnFDI(2 lags)</i>	-	0.0788* (0.0425)	-		
<i>LnFDI*AC</i>	-	-	-0.0268*** (0.00822)	-0.0382*** (0.0110)	0.0697*** (0.0227)
<i>LnAGE</i>	-	-	-		-3.330*** (0.516)
<i>LnSCHOL</i>	-	-	-		-0.0853*** (0.0240)
<i>Constant</i>	-0.706*** (0.0529)	-0.616*** (0.0158)	-0.619*** (0.0584)	-0.562*** (0.0726)	11.65*** (1.767)
AIC	266.0159	83.0699	258.6431	265.9425	268.071
BIC	276.8112	97.4839	272.9317	280.2615	289.5494
Observations	268	130	259	263	263

Robust standard errors in parentheses. *** p<0.01; ** p<0.05; * p<0.1.

of this study, which proposes that FDI moderated by AC has a more significant effect than FDI without being moderate. However, note that in both models the isolated FDI impact was higher than in the previous models, respectively, 20% and 22.6%.

However, in the analysis of the Information Criteria of models 3 and 4, model 3 (AIC = 258.6431 and BIC = 272.9317), when compared to model 4 (AIC = 265.5425 and BIC = 280.2615), explains the relationship between the FDI moderated by AC with the Malmquist Index. Thus, it is possible to infer that the variable number of workers with doctorates presents themselves as CA's best proxy. In companies, the doctoral professional can help entrepreneurs gain insights from data, assisting in better business decisions and predicting trends, since they position themselves strategically among competitors.

It is worth mentioning that AC measurement is very complex because it deals with stored knowledge and, therefore, other variables may be interesting to measure it, given the lack of availability of data for the analyzed years.

The negative and statistically significant sign of the interaction between FDI and AC shows that the greater the number of professionals with doctorates or R&D workers combined with the inflow of foreign capital generates lower productivity. Some possible explanations for this result are: 1 - due to the labor costs of this type of professionals, which does not justify the productivity generated; 2 - the innovation generated by R&D professionals have not justified the productivity found in Brazilian municipalities; and, 3 - professionals with higher education have not generated the expected productivity in the economy.

To solve this problem, Model 5 isolated some characteristics of the workers. The experiment was used as an explanatory variable of productivity. The proxy used for the experiment was the age of the workers. We also isolate the effect of education on productivity by using the number of R&D workers as proxy for Absorptive Capacity.

For model 5, it can be stated that both the mean age and the number of professionals with higher education presented with high statistical significance (1%), but with an opposite sign than expected, which does not prove the hypotheses H₄ and H₅.

The average age of workers, used as a proxy for the workers' experience, showed a negative correlation, that is, an increase in the work experience does not positively affect productivity. Thus, other worker factors can be analyzed such as employee satisfaction, organizational commitment and management skill. It may also be due to the fact that workers with a higher average age have greater difficulties in absorbing new technologies, which negatively affects the productivity of municipalities.

Regarding the educational level, the estimates showed that the number of workers with higher education has no significant effect on the increase of the Malmquist index. This can be explained by the fact that there are few professionals in the area of Exact Sciences in the country, which results in the market demand for the

training of professionals with qualification fundamentally in the areas of Engineering, Physics, Statistics, among others related.

It is emphasized that the inadequacy of the workforce in relation to the functions is among the causes of a structural problem of the country whose solution is essential for the resumption of growth: low labor productivity. The indicator that measures the efficiency of a country's labor force is influenced by factors such as the business environment and the investment in the improvement of productive processes, aspects in which Brazil does not present good results.

In Model 5 it was possible to confirm Hypothesis 3. The interaction between FDI and AC is positive and statistically significant at the 1% level. In addition to the direct effect of FDI (10.8%) and Economic Complexity (1.06%), there is an interaction effect between FDI and AC (6.97%). This means that the greater the number of workers in R&D and Foreign Direct Investment, the higher the productivity of the municipality.

As a policy recommendation, the interaction effect shows that external capital inflows remain the most important variable to increase productivity. However, it is necessary to ensure that there is AC in place in order to increase the productivity of the region. Economic Complexity, FDI and AC together tend to raise productivity by approximately 19%.

6. Final considerations

This article aimed to analyze the impact of FDI and AC on Productivity of 106 Brazilian municipalities between 2010 and 2014, whose productivity was measured by the Malmquist Index, which is characterized by the ability to measure change, in terms of Total Productivity of Factors, between different periods and to decompose this index into technical efficiency and change of technology.

Initially, the positive effect of Economic Complexity and Foreign Direct Investment on productivity was discussed. Understanding that Foreign Direct Investment affects the economy is of great importance for public policy planning and for the efficient allocation of government efforts and resources. Thus, the importance of adopting FDI policies to ensure the continuity of productivity growth and the diffusion of new practices and technologies is highlighted.

However, it is worth mentioning that the FDI lag model points to a maturation period so that the inflow of financial resources from abroad has a significant impact on the productivity of the municipalities. This is because certain investments, such as high technology research projects, require more time to get returns.

Finally, by the analysis, it was verified that the experience of the workers (proxy of the average age of workers) does not positively affect productivity. Thus, it is suggested for future work the inquiry of other factors of the worker that can impact productivity, such as employee satisfaction, organizational commitment and management skills.

As for the level of schooling, estimates have shown that the number of workers with higher education has no significant effect on the increase in the Malmquist index. This can be explained by the fact that there are few professionals in the area of Exact Sciences in the country, which results in the market demand for the training of professionals with qualification fundamentally in this area.

Therefore, FDI moderated by AC proxies was analyzed, although they presented high statistical relevance and did not provide the expected signal (Models 3 and 4). Nevertheless, by the Information Criterion the variable number of professionals with a doctoral degree presented itself as the best proxy of the AC, highlighting the importance of testing other variables as a proxy to the AC.

However, Model 5 showed that after controlling for characteristics of the workers (experience and education) it was possible to find a positive and statistically significant effect for the interaction between FDI and AC. Thus, one of the limitations of this paper was the lack of data availability for AC, such as investments in R&D, a variable that is more commonly used in the literature. It is suggested as future studies the use of other variables for AC in order to compare the results and also to test endogeneity problems.

From the results found, we conclude that the Economic Complexity had a smaller effect on productivity compared to the FDI, which always shown a positive and statistically significant impact on productivity, especially when combined with Absorptive Capacity.

In order to increase productivity in Brazilian municipalities, it is necessary to encourage policies for the inflow of foreign capital in order to generate new business, markets and products. However, professionals are needed that are able to absorb the new knowledge generated, in order to generate a greater impact on Brazilian productivity.

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Annex 1

ALVORADA	MAGE
APARECIDA DE GOIANIA	MANAUS
ARACAJU	MARACANAU
ARARAS	MATA DE SAO JOAO
BARRETOS	MIRACEMA DO TOCANTINS
BARUERI	MOGI DAS CRUZES
BAURU	MOGI MIRIM
BELO HORIZONTE	MOGI-GUACU
BLUMENAU	MONTE CARMELO
BRASILIA	Montes Claros
BRUSQUE	NATAL
CAMACARI	NITEROI
CAMPINAS	NOVA FRIBURGO
CAMPO GRANDE	NOVA LIMA
CANOAS	NOVA ODESSA
CASCAVEL	OSASCO
CATAGUASES	PALMAS
CAXIAS DO SUL	PASSO FUNDO
CONTAGEM	POJUCA
CORNELIO PROCOPIO	PONTA GROSSA
COTIA	PORTO ALEGRE
CRAVINHOS	PRADOPOLIS
CRUZEIRO	QUATRO BARRAS
CUIABA	RECIFE
CURITIBA	REGISTRO
DIADEMA	RIBEIRAO PRETO
DUQUE DE CAXIAS	RIO DE JANEIRO
ELDORADO DO SUL	RIO DO SUL
EMBU DAS ARTES	RIO NEGRO
EUSEBIO	RONDONOPOLIS
FLORIANOPOLIS	SALTO
FORTALEZA	SALVADOR
FRAIBURGO	SANTA BARBARA D'OESTE
FRANCA	SANTA CRUZ DO SUL
GOIANIA	SANTA ISABEL
GRAVATAI	SANTO ANDRE
GUARULHOS	SAO BERNARDO DO CAMPO
IGARASSU	SAO CAETANO DO SUL
IMBITUBA	SAO JOSE DO RIO PRETO
INDAIA TUBA	SAO JOSE DOS CAMPOS
ITAJAI	SAO JOSE DOS PINHAIS
ITAQUAQUECETUBA	SAO LEOPOLDO
ITATIBA	SAO LUIS
ITIRAPINA	SAO SEBASTIAO DO CAI
ITU	SAPUCAIA DO SUL
ITUPEVA	SERRA
JARAGUA DO SUL	SERTAOZINHO
JOAO PESSOA	SOBRAL
JOINVILLE	TIJUCAS
JUNDIAI	TIMBO
LAJEADO	TRES LAGOAS
LINS	UBERABA
LONDRINA	UBERLANDIA