

## Article

# Environmental Kuznets Curve and the Pollution-Halo/Haven Hypotheses: An Investigation in Brazilian Municipalities

Eduardo Polloni-Silva <sup>1,\*</sup> , Diogo Ferraz <sup>2,3,4</sup> , Flávia de Castro Camioto <sup>5</sup>, Daisy Aparecida do Nascimento Rebelatto <sup>6</sup> and Herick Fernando Moraes <sup>1</sup>

<sup>1</sup> Department of Production Engineering (DEP), Federal University of São Carlos (UFSCar), São Carlos 13565-905, Brazil; herickmoraes@dep.ufscar.br

<sup>2</sup> Department of Economics (DEECO), Federal University of Ouro Preto (UFOP), Mariana 35420-000, Brazil; diogoferraz@alumni.usp.br

<sup>3</sup> Department of Production Engineering, São Paulo State University (UNESP), Bauru 17033-360, Brazil

<sup>4</sup> Department of Innovation Economics, University of Hohenheim, 70599 Stuttgart, Germany

<sup>5</sup> Department of Production Engineering, Federal University of Triângulo Mineiro (UFTM), Uberaba 38025-180, Brazil; flavia.camioto@uftm.edu.br

<sup>6</sup> Department of Production Engineering, Escola de Engenharia de São Carlos (EESC), University of São Paulo (USP), São Carlos 13566-590, Brazil; daisy@usp.br

\* Correspondence: eduardopolloni1@gmail.com; Tel.: +55-16-3351-9509



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**Abstract:** There is much discussion on the non-linear relationship between economic growth and carbon dioxide (CO<sub>2</sub>) emissions. Additionally, the effects of Foreign Direct Investment (FDI) on the environment are ambiguous, as both beneficial (i.e., pollution-halo) and harmful (i.e., pollution-haven) effects were found. Therefore, the literature presents no consensus on either of these topics. This is especially problematic for developing regions, as these regions represent growing economies interested in receiving foreign investments, and their CO<sub>2</sub>-related research is limited. This study aims to understand the impacts of economic growth and FDI on the CO<sub>2</sub> emissions of São Paulo state, Brazil. To perform this study, a unique dataset on regional FDI was built, and 592 municipalities were included. The analyses combine linear and non-linear estimations, and the results suggest a non-linear relationship between Gross Domestic Product (GDP) per capita and CO<sub>2</sub> emissions, along with a negative association between FDI and CO<sub>2</sub>. Finally, this study discusses possible policy implications and contributes to the international literature.

**Keywords:** foreign direct investment; FDI; economic growth; CO<sub>2</sub> emissions; pollution; São Paulo; Brazil

## 1. Introduction

Economic growth plays a vital role in countries' economic and human development. Over the past 40 years, the world economy has more than tripled [1]. Despite economic growth raising living standards in most countries, it was also responsible for an increase in CO<sub>2</sub> emissions and a depletion of natural resources. The results presented in the IPCC report confirm that global warming is largely attributed to human activities, especially those related to CO<sub>2</sub> emissions [2].

According to Mardani et al. [3], the majority of previous studies for the past two decades have been intensively focused on the study of the nexus between economic growth and energy use and their effects on CO<sub>2</sub> emissions.

Dinda [4] and Liu et al. [5] states that the investigation of the relationship between economic growth and CO<sub>2</sub> emissions mainly focuses on whether this relationship follows the Environmental Kuznets Curve (EKC) hypothesis. According to this hypothesis, the environmental quality deteriorates in the early stage of economic development/growth and improves in a later stage as the economy develops.

Most studies following this approach confirm the existence of an inverted U-shaped (and other curve shapes) for the relationship between growth and emissions (i.e., EKC) in

many countries and regions [6–9]. Still, Al-mulali et al. [10] and Sadik-Zada and Ferrari [11] state that this transition in the GDP–CO<sub>2</sub> relationship depends on the region's economic development stage and technology adoption, which arguably explains how the EKC theory still generates debate across distinct regions/countries.

Accordingly, pollutant emissions are the most critical environmental problem world-wide, and their reduction is part of the political agenda and international organizations debate, such as the Paris Agreement and the 2030 Agenda for Sustainable Development [12]. According to the Sustainable Development Goals (SDG), countries and local governments must foster climate action (SDG-13) and global partnerships (SDG-17), as well as inclusive industrialization, innovation, and economic growth (SDG-8 and SDG-9) to achieve a better stage of sustainable development.

Moreover, policies must be adequate to deal with the reduction of CO<sub>2</sub> emissions without affecting economic growth, especially in developing countries, which have been stressing efforts to promote and upgrade their industrial activity [13,14]. Generally, productivity growth is associated with growth in real wages, and ultimately an improvement in living standards [14].

In this sense, Foreign Direct Investment (FDI) is a remarkable phenomenon for both developed and developing countries due to its notorious effects on globalization [15] and economic growth [16]. Bakhsh et al. [17] state that FDI contributes to the host economy by stimulating its economic development, by being a source of external finance, and by reducing the bridge between domestic savings and target investment. In sum, FDI facilitates local development, which further encourages employment opportunities for skilled and unskilled labor in the host country [17].

However, there is still some debate on the effects of FDI on carbon emissions. In sum, a multitude of studies argues FDI may negatively affect the environment due to its impact on economic growth [7,8,16,18,19]. While FDI inflows may rapidly affect the host economy and bring hope to the deprived populations, they may also increase environmental degradation and pollution [16]. In other terms, the promotion of economic growth by inward FDI is not free from environmental costs [20], which is alarming considering that nations, especially developing countries, are eager to attract foreign investments [21]. Indeed, He and Yao [16] assert that emerging countries may trade off FDI's financial benefits against environmental losses.

Nonetheless, other studies pointed out that FDI may bring green technologies, which could improve the host country's environmental quality and boost productivity [17,21]. Also, foreign investments might create new investments in the energy sector, promoting alternative energy resources such as wind turbines and photovoltaic cells [22]. These cleaner resources can reduce pollution caused by coal and diesel consumption. Moreover, Shahbaz et al. [20] argue that the effect of FDI on the environment depends on the group of analyzed countries. For example, Yi et al. [23] found that green FDI has significantly reduced China's carbon emissions.

In addition to the current lack of consensus on the FDI–CO<sub>2</sub> relationship, a major limitation is the lack of regional studies. Research on FDI usually employs country-level data, with a few exceptions. For instance, He and Yao [16] studied FDI in 29 Chinese provinces to avoid significant regional disparities across the country. A similar approach was presented by Hille et al. [24] and their research in Korea. Zhang and Zhou [25] also argue that FDI studies rarely consider regional differences, although these differences may alter results. In fact, the international business literature recognizes the importance of regional or "subnational" differences in FDI activities [26,27], and claim that these subnational differences are particularly relevant for large emerging economies like the BRICS [28]. Specifically, Brazil's inequality and regional differences are well documented [29].

To fill this gap, our study aims to analyze the impact of inward FDI on CO<sub>2</sub> emissions in Sao Paulo state, Brazil. We employed advanced econometric models and a unique panel dataset regarding FDI within 592 municipalities in São Paulo state ranging from 2010 to 2016. This dataset is the first and only regional FDI database available for Brazil, which

implies an extensive manual check of approximately 22,000 export records to identify the source of capital of a company in a given municipality. Such data allows an in-depth analysis of FDI in the region and may serve as a base for future research worldwide.

We choose Sao Paulo for three reasons. First, Brazil has received little attention from the international literature. The absence of empirical studies in Brazil is particularly worrisome considering the country's inward FDI stock. The Sao Paulo state is particularly relevant for a more in-depth analysis of FDI in Brazil due to foreign investments in the region. According to the Central Bank of Brazil [30], the state accounted for roughly 39% of foreign investments in the country, at least considering industrial enterprises. Besides, Sao Paulo presented the highest Gross Domestic Product (GDP) in 2016 and is the richest and most economically complex region in Brazil [31]. Second, Brazil is the leading air polluter in Latin America and is a country that experienced significant growth in recent years [32,33], with 441.8 million tons of CO<sub>2</sub> emissions in 2018 [34]. Sao Paulo state is responsible for  $71.50 \times 10^6$  tCO<sub>2</sub>/year [35], which is remarkable to control environmental damage in the country. In addition, Brazilian policymakers do not fully understand the environmental impacts of such investments. Third, there is an absence of studies using subnational data for developing countries.

Accordingly, our study presents three contributions to the literature. First, we present an in-depth CO<sub>2</sub>-related analysis in a developing region, which is important for local policymaking, especially for policymakers being pressured to promote economic growth with minimum environmental damage. Policymakers in other parts of Brazil and the world may access our policy recommendations. Second, due to the scarcity of data, we built a unique dataset with information about multinational companies located in Sao Paulo state using an export-oriented FDI measure similar to Morales and Moreno [36] and Polloni-Silva et al. [37]. Third, our findings reveal the impact of FDI on CO<sub>2</sub> emissions at a municipal level, therefore contributing to the international debate on this issue.

The rest of this paper is organized as follows. The literature review on the impacts of FDI and the possible non-linear relationship between economic growth, FDI, and emissions is presented in Section 2. Section 3 focuses on the data gathering process and the estimation strategy. Section 4 presents the main findings. Section 5 presents the robustness checks. Finally, Section 6 presents the policy implications of our findings, along with a brief conclusion.

## 2. Literature Review and Hypotheses Development

### 2.1. Pollution-Halo and Pollution-Haven

According to the literature, FDI may boost the host country's economic growth while putting the environment at risk [16,20,38]. Although previous research has found inconclusive results, scholars have attempted to explain the influence of FDI on the environment, as it is often assumed to be connected to economic growth [39,40].

Nonetheless, the literature points to two alternative theories that associate FDI and the environment. Namely, the "pollution-haven" and "pollution-halo" theories. The pollution-haven hypothesis defends that inward FDI is attracted to weak regulatory and environmental laws presented by the host country [17,41]. Thus, FDI should negatively affect the environment, especially in developing economies with weaker environmental standards [42]. In fact, several countries are eager to attract FDI [21] and may use their lax environmental regulations to compete in the "FDI tournament" [38]. On the contrary, the pollution-halo hypothesis claims that foreign firms, especially from developed economies, arrive with newer and cleaner technologies and better management practices capable of improving the environmental quality of the host country [17,21]. The pollution-halo hypothesis is especially relevant when domestic firms are considered pollution-intensive when compared to international standards [17].

Concerning the pollution-haven and pollution-halo studies, the results are debatable, once again. Zafar et al. [21] found evidence of FDI bringing advanced technology and innovative products in the United States, which decreased environmental pollution as the

pollution-halo theory suggests. Additionally, Jiang et al. [43] claim that FDI improved air quality in China, especially in populated cities.

By employing cointegration analysis, Nepal et al. [44] found that an increase in FDI inflows may reduce energy use, evidencing that energy-efficient techniques through FDI can reduce carbon emissions, which reinforces the importance of policies that encourage FDI inflow in renewable energy sectors. Evidence of pollution-halo in developing economies is also found by Tiba and Belaid [15] and Yi et al. [23].

However, FDI may present a significant positive impact on environmental degradation, thus supporting (especially in developing economies) the pollution-haven hypothesis. In this sense, He and Yao [16] found (at least to some extent) evidence to pollution-havens in China, while similar results are also presented by Jiang [8], Shahbaz et al. [20], and Opoku and Boachie [19]. Furthermore, Kastratović [42] studied 63 different developing countries, including Brazil, and found evidence of pollution-havens. Zakarya et al. [45] studied the BRICS and also found evidence of pollution-havens, as FDI increased CO<sub>2</sub> emissions in the selected countries.

Yet, the results presented by Shao et al. [46] refuted the pollution-haven hypothesis for the BRICS. Thus, in essence, the literature still lacks conclusive results. As Nasir et al. [18] declared, FDI may have either a positive or negative impact on the host's environment. It depends on the research method, timeframe, and region [16,18,25]. *De facto*, inconclusive or mixed results for developing countries were also found by Nadeem et al. [47], Chen and Yang [48], and Xu et al. [49]. Therefore, further research is essential to better comprehend the link between FDI and the environment.

Thus, considering the current literature on FDI, environment, and location choice, we formulate the following hypothesis:

**Hypothesis 1 (H1).** *The effect of FDI on CO<sub>2</sub> emissions is statistically significant.*

Which can be partitioned into:

**Hypothesis 1a (H1a).** *FDI is negatively associated with CO<sub>2</sub> emissions (pollution-halo behavior);*

**Hypothesis 1b (H1b).** *FDI is positively associated with CO<sub>2</sub> emissions (pollution-haven behavior).*

## 2.2. Non-Linear Relationship between Economic Growth, FDI, and CO<sub>2</sub> Emissions

Global warming is one of the main concerns of world globalization in terms of industrial development. There has been a steady growth in energy demands to sustain the development of countries. At the same time, there are increasing concentrations of greenhouse gases (GHGs) in the atmosphere, particularly carbon dioxide (CO<sub>2</sub>), intensifying the natural greenhouse effect and the heat it generates. According to IPCC [2], human activities are estimated to have caused approximately 1.0 °C of global warming above preindustrial levels.

Since the 1990s, some authors have come to defend that the interaction between economic growth and environmental problems has a certain regularity. This empirical evidence first emerged by Grossman and Krueger [50,51], in which the relationship between GDP per capita and emission of pollutants takes the form of a "U-inverted" or Environmental Kuznets Curve (EKC). Therefore, their work linked economic growth to environmental degradation in a non-linear way.

The EKC theory asserts that environmental degradation increases in the early stages of economic growth, and decreases when the income surpasses a certain threshold level [16,50]. In other words, the EKC theory suggests that at an early stage of economic development, people will demand more energy and natural resources, while neglecting the environment. However, at a later stage of development, people may present a different mindset and demand a cleaner environment, energy-efficient products, and the protection of natural resources, as well as the environment, in general [18,21]. In summary, EKC represents the

notion that despite the environmental degradation at first, economic growth can improve the environmental quality in the long-run [18]. As FDI is often assumed to be connected to economic growth [39,40], multiple scholars are now analyzing FDI and its effects on the environment from the EKC perspective.

However, results are still inconclusive [18]. Both country-specific and diversified studies have been published. Fosten et al. [52] found significant evidence of EKC in the UK. Esteve and Tamarit [53] studied Spain and found that, in the long-run, economic growth reduces CO<sub>2</sub> emissions. However, Ghosh [54] found no conclusive evidence of this long-run relationship between growth and emissions in India, and similar results were found for Turkey [55], where pollution was not found to be associated with FDI or economic growth. Instead, Haug and Ucal [56] examines the effects of foreign trade and FDI on CO<sub>2</sub> emissions in Turkey and the results show that FDI has no statistically significant long-run effects and CO<sub>2</sub> intensity is not influenced by FDI, but an environmental Kuznets curve is present for CO<sub>2</sub>.

More recently, Nasir et al. [18] found evidence of EKC in Asia, although their results were not statistically significant. Contrarily, the authors of [57] validated the EKC phenomenon in Malaysia. Furthermore, Onafowora and Owoye [58] analyzed the EKC phenomenon in selected economies such as Brazil, Mexico, China, Japan, and South Korea (among others) and only found evidence of EKC in developed economies. Malik et al. [59] studied the case of Pakistan and confirms the EKC hypothesis, and the symmetric results show that economic growth and FDI intensify carbon emission in both the long and short-run and Pakistan is a pollution haven for FDI.

The results of Liu et al. [5] revealed that there is a reverse U-shape relationship between economic growth and carbon dioxide emissions in China, and an inverted N-shaped relationship between FDI and carbon dioxide emissions. However, he suggested reexamining the results by grouping different regions.

Considering the literature, it is possible to infer that the results on the non-linear relationship between economic growth and environmental degradation are still inconclusive [16].

Considering the current literature on economic growth, EKC, and FDI, the following hypotheses were formulated:

**Hypothesis 2 (H2).** *Economic growth presents a non-linear relationship with CO<sub>2</sub> emissions;*

**Hypothesis 3 (H3).** *FDI presents a non-linear relationship with CO<sub>2</sub> emissions.*

### 3. Data and Method

Our sample includes 592 out of the 645 municipalities in São Paulo state. We excluded 1 municipality for the lack of industry-related data and 52 small-sized municipalities (i.e., averaging 3000 inhabitants each) for the lack of energy-related data. Consequently, the employed sample comprises roughly 99.3% of the state's population. Also, the data ranges from 2010 to 2016, resulting in a seven-year period and 4144 observations.

#### 3.1. CO<sub>2</sub> Emissions

This study employs CO<sub>2</sub> emissions per capita as the dependent variable. However, a detailed (i.e., municipal-level) CO<sub>2</sub> database is not available at this moment. Previous scholars also faced this issue when studying CO<sub>2</sub> at a more micro-level. For this reason, we manually calculate the emissions from energy consumption for the municipalities of São Paulo, which is a common procedure in CO<sub>2</sub>-related studies.

Therefore, we use data on electricity consumption and fossil fuels. Namely, the calculation includes gasoline, diesel, liquefied petroleum gas, fuel oil, aviation gasoline, and aviation kerosene. By following this calculation, this study is aligned with previous research, employing data from the Brazilian Energy Research Office (EPE) and the National Agency for Petroleum, Natural Gas and Biofuels (ANP). Yet, the Brazilian fuel legislature



makes it inappropriate to calculate CO<sub>2</sub> emission directly. Thus, this study followed the IPCC [60] guidelines and subtracted the volume of ethanol added to gasoline C, which is commonly commercialized in the country, to use gasoline A in our calculations. The volume of biodiesel added to diesel oil was also subtracted.

Finally, the CO<sub>2</sub> emissions are calculated using the conversion factors presented in Table 1.

**Table 1.** CO<sub>2</sub> conversion factors for various energy sources.

Type of Energy	Conversion Factor	Description
Electricity *	0.0512 (2010); 0.0292 (2011); 0.0653 (2012); 0.096 (2013); 0.1355 (2014); 0.1244 (2015); 0.0817 (2016)	tCO <sub>2</sub> /Mwh
Automotive gasoline	69.3	tCO <sub>2</sub> /TJ
Diesel oil	74.1	tCO <sub>2</sub> /TJ
Liquefied petroleum gas	63.1	tCO <sub>2</sub> /TJ
Fuel oil	77.4	tCO <sub>2</sub> /TJ
Aviation gasoline	70	tCO <sub>2</sub> /TJ
Aviation kerosene	71.5	tCO <sub>2</sub> /TJ

Source: Ministry of Science, Technology, Innovation, and Communications (MCTIC) and IPCC (2006). \* The Brazilian government considers all electricity sources (i.e., hydro, natural gas, oil, biomass, coal, wind, nuclear, and solar) contributing to the national grid to define the CO<sub>2</sub> conversion factor.

### 3.2. Explanatory and Control Variables

This study is interested in better understanding the general effects of the development of São Paulo state on the environment. Accordingly, the regions' GDP per capita is employed as a measure for economic growth, as done by previous CO<sub>2</sub>-studies [7,61,62]. Although alternative variables (e.g., income levels) could provide valuable information on the overall richness of a region, there is a general consensus on the importance of GDP as the primary measure of how an economy or region is performing.

Additionally, in recent years, FDI received considerable recognition in many CO<sub>2</sub>-related studies [5,15,62]. However, similar to the CO<sub>2</sub>-issue presented earlier, São Paulo state does not present a regional FDI database. De facto, there are no governmental statistics on FDI at the state or city-levels [36], and this lack of data virtually explains the current absence of Brazilian studies on FDI and its regional effects, except for Morales and Moreno [36] and Polloni-Silva et al. [37].

Therefore, we build a proxy for the regional FDI presence in each municipality using export-data from the Brazilian Integrated System of Foreign Trade (SISCOMEX). This system provides information on all exporting companies operating in São Paulo state, including their names, addresses, and export band value (e.g., up to 1 million BRL, between 1 and 5 million BRL, between 5 and 10 million BRL). With this information, a research group manually verified if these companies presented international origins (i.e., if the headquarters is located in a foreign country). For the whole São Paulo state during the 2010–2016 period, the research group performed approximately 22,000 individual checks. Hence, we follow Morales and Moreno [36] in building a proxy variable for the intensity of FDI in a region, following Equations (1) and (2) below.

$$FDI_{jt} = W_{jt} \times \left( \frac{MNC_{jt}}{TC_{jt}} \right) \quad (1)$$

And,

$$W_{jt} = \frac{SFE_{jt}}{STE_{jt}} \quad (2)$$

where *FDI* is the intensity of FDI for every region *j* at the year *t*, *W* is the adjustment weight regarding the exports from foreign companies, *MNC* is the number of multinational companies (i.e., foreign companies), *TC* is the total number of exporting companies, *k* is the number of foreign companies, *SFE* is the sum of exports from foreign companies, and *STE*

is the sum of total exports (including national and foreign companies). The export band values (i.e., 1 million; 5 million; 10 million; 50 million, 100 million, and 320 million) are used to calculate *SFE* and *STE*, similar to Morales and Moreno [36].

This FDI measure varies from 0 (no foreign companies) to 1 (all exports are dependent upon foreign companies). Thus, this measure represents the intensity of FDI for every municipality in every year of the sample and shows the relative importance of FDI for each region. With this information, our study will be the first to evaluate the municipal-level effects of FDI on the environment in a Brazilian region.

It should be noted that this FDI measure is not perfect as it excludes foreign companies commercializing their products exclusively within the country. However, this measure is arguably a valuable proxy for FDI considering that many foreign companies use Brazil as an export station to other Mercosul (*Mercado Comum do Sul*) countries [63]. As an example, roughly 22% of Paraguay's imports and 24% of Argentina's imports come from Brazil, and these products involve a multitude of industrial sectors (e.g., machinery, chemicals, clothing, cars) [64]. As the SISCOMEX system presents data on all exporting companies from all sectors, our FDI measure is arguably capable of including a considerable share of all foreign companies in the country. Additionally, a similar export-oriented proxy for FDI was already employed by Morales and Moreno [36] and Polloni-Silva et al. [37].

Furthermore, regarding the control variables, previous scholars have employed a multitude of demographic, economic, and technology-related variables in previous studies. For this study, the models include the industrial and service sectors' share of total GDP, the share of electricity consumption in total consumption, and the population density. Therefore, our model is similar to previous employed models [9,61,62,65]. Table 2 presents the descriptive statistics and the sources used.

**Table 2.** Descriptive statistics and variables description.

Variable	Description	Source	Observations	Mean	S.D.	Min.	Max.
CO <sub>2</sub>	CO <sub>2</sub> per capita (tons)	-	4144	2.055	2.540	0.105	34.413
GDPPC	GDP per capita (2010 values; given in 1000 BRL)	Brazilian Institute of Geography and Statistics (IBGE)	4144	20.197	16.578	4.780	278.840
FDI	Foreign Direct Investment	-	4144	0.046	0.144	0.000	1.000
IND	Industrial sector (value added to total GDP; %)	IBGE	4144	19.365	13.303	2.224	85.966
SERV	Service sector (value added to total GDP; %)	IBGE	4144	39.147	11.086	7.412	75.640
RES	Share of residential electricity consumption in total consumption (%)	Secretary of Energy and Mining of the State of São Paulo	4144	34.679	12.072	0.177	66.417
DENS	Population density (population per square kilometer)	IBGE	4144	342.577	1301.545	3.698	13,534.820

### 3.3. Explanatory and Control Variables

We start the analysis by verifying the linear effects of the explanatory variables on the regions' CO<sub>2</sub> emissions. Thus, the primary model can be written as:

$$\ln CO_{2jt} = \beta_0 + \beta_1 \ln GDPPC_{jt} + \beta_2 FDI_{jt} + \beta_3 \ln' X_{jt} + a_j + e_{jt} \quad (3)$$

where CO<sub>2</sub> is the dependent variable for every municipality *j* at time *t*, GDPPC is the real GDP per capita, FDI is the aforementioned proxy for regional FDI intensity, X is the set of control variables, *a<sub>j</sub>* is the regional fixed effect, and *e* is the error term. All variables are employed in their natural-log form, except for FDI as the variable is already in percentage form. In addition to the linear effects of the other explanatory variables, this model formulation allows us to verify if FDI presents a pollution-haven or pollution-halo behavior.

However, this study also aims to test for possible non-linear relationships between economic growth, FDI, and CO<sub>2</sub> emissions. Thus, by following many previous scholars, we add quadratic terms to our model:

$$\ln CO_{2jt} = \beta_0 + \beta_1 \ln GDP_{PCjt} + \beta_2 \ln GDP_{PCjt}^2 + \beta_3 FDI_{jt} + \beta_4 \ln' X_{jt} + a_j + e_{jt} \quad (4)$$

And,

$$\ln CO_{2jt} = \beta_0 + \beta_1 \ln GDP_{PCjt} + \beta_2 FDI_{jt} + \beta_3 FDI_{jt}^2 + \beta_4 \ln' X_{jt} + a_j + e_{jt} \quad (5)$$

with these models, it is possible to test for U or inverted-U curves between GDP, FDI, and the CO<sub>2</sub> emissions, thus verifying the existence of EKC curves and better understanding how future development may affect the environment if São Paulo continues to grow and attract foreign investments.

Furthermore, it is necessary to consider possible non-spherical disturbances that are commonly found in panel datasets. Initially, we verified if our baseline models present collinearity issues (see Appendix A), which is a common issue in environmental studies [66]. Moreover, the Hausman test was employed to examine which specification (i.e., random or fixed-effects) was better suited for our models. Additionally, the modified Wald test for group-wise heteroscedasticity [67], the Wooldridge test for autocorrelation [68], and the Pesaran test for cross-sectional dependence [69] were used. In sum, these tests' results show that the fixed-effects model was the preferable specification and that the sample shows heteroscedasticity, autocorrelation, and cross-sectional dependence problems.

To deal with these issues, CO<sub>2</sub>-studies have been using the Driskoll–Kraay [70] estimator [9,61,71], as this technique deals with heteroscedasticity and autocorrelation problems and is robust to general forms of cross-sectional (spatial) and temporal dependence with panels of all sizes, including an N > T dataset [72]. Therefore, the fixed-effects Driscoll–Kraay (DK) estimator is the main technique employed in the analyses. Yet, the sample is limited as a higher time frame is not available at the moment. Thus, to take advantage of the higher degrees of freedom as the random-effects specification has, the Feasible Generalized Least Squares (FGLS) technique was employed to double-check our findings. FGLS is usually a random-effects estimator capable of dealing with heteroscedasticity, autocorrelation, and cross-sectional correlation [68,73].

#### 4. Main Results

Our main findings are presented in Table 3, which shows the linear and quadratic relationships between our explanatory variables and the regions' CO<sub>2</sub> emissions. As shown, economic growth increases CO<sub>2</sub> emissions, as expected. Moreover, Table 3 shows that a non-linear relationship between GDP per capita and CO<sub>2</sub> emissions exists, as the quadratic term for GDP per capita is significant and negative. Thus, Table 3 suggests that after a certain threshold, the overall growth/richness of a region, here represented by the GDP per capita, helps to decrease the emission level.

Additionally, FDI has a linear and negative relationship with CO<sub>2</sub> emissions, which indicates that FDI presents a pollution-halo behavior. Thus, these results are aligned with some previous studies defending the pollution-halo hypothesis for FDI such as Jiang et al. [74], Zhang and Zhou [25], and Shao et al. [46]. Indeed, FDI seems to be environmentally beneficial to the region. Furthermore, a non-linear relationship between FDI and CO<sub>2</sub> is not found, as the quadratic term is not significant. Yet, before declaring support for any hypothesis, further tests will be made.



**Table 3.** DK and FGLS estimations for the whole sample.

Variables	(1)	(2)	(3)	(4)	(5)	(6)
	DK	FGLS	DK	FGLS	DK	FGLS
GDPPC	0.309 *** (0.0251)	1.372 ** (0.685)	0.901 *** (0.120)	5.518 ** (2.457)	0.308 *** (0.0252)	0.913 *** (0.306)
GDPPC <sup>2</sup>			−0.0916 *** (0.0164)	−0.729 ** (0.348)		
FDI	−0.0614 * (0.0336)	−2.563 ** (1.265)	−0.0553 * (0.0319)	−0.636 (1.177)	−0.00806 (0.0975)	−2.278 (9.335)
FDI <sup>2</sup>					−0.0551 (0.0762)	0.324 (9.171)
IND	0.0458 *** (0.0138)	1.340 ** (0.535)	0.0427 ** (0.0182)	0.607 (0.477)	0.0459 *** (0.0137)	1.721 *** (0.591)
SERV	0.503 *** (0.0872)	2.033 ** (0.859)	0.489 *** (0.0985)	−0.584 (1.020)	0.503 *** (0.0871)	1.662 ** (0.784)
RES	−0.221 *** (0.0381)	−0.250 (0.364)	−0.218 *** (0.0390)	−0.152 (0.328)	−0.221 *** (0.0380)	1.383 (1.409)
DENS	0.990 ** (0.401)	−1.376 * (0.778)	1.048 *** (0.403)	−0.0203 (0.736)	0.988 ** (0.401)	−0.947 (0.638)
Constant	−5.800 *** (2.130)	−7.787 ** (3.648)	−6.909 *** (2.270)	−7.523 * (4.055)	−5.792 *** (2.128)	−13.43 * (7.236)
F	228.10 ***		193.48 ***		728.43 ***	
Wald Chi2		33.97 ***		27.99 ***		41.45 ***
Hausman	117.64 ***		200.59 ***		114.69 ***	
Mod. Wald.	190,000 ***		200,000 ***		190,000 ***	
Wooldridge	343.60 ***		328.73 ***		340.18 ***	
Pesaran	294.31 ***		290.91 ***		294.45 ***	
Observations	4144	4144	4144	4144	4144	4144
Municipalities	592	592	592	592	592	592

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Standard errors in parentheses. DK: Fixed-effects regression with Driscoll–Kraay standard errors. FGLS: Feasible Generalized Least Squares.

Regarding the control variables, Table 3 shows that both the industrial and the service sectors contribute to higher emission levels. Again, this demonstrates that economic growth is environmentally costly. The results demonstrate that the industrial sector is positive and significant, while FDI is negative and significant. Thus, the overall industrial sector increases the emission levels, while a higher FDI intensity is beneficial.

Likewise, the service sector, which is ignored by many scholars, also increases pollution. This result is expected considering that the CO<sub>2</sub> measure employed by our study is calculated using energy consumption. De facto, the service sector is expanding in the region, and many services, particularly in highly urbanized regions, involve road transportation, which generates higher levels of pollution [75]. In addition, a higher share of residential electricity consumption in the total consumption presents negative coefficients, indicating that higher shares of consumption by other sectors (e.g., industry) result in higher CO<sub>2</sub> than the residential sector. In other words, the impact from the residential sector is lower than other sectors. Also, regions with higher population density levels present higher emissions, as these regions are arguably larger and more populated.

To further evaluate if the results from Table 3 are reliable, we split the sample into subsamples. First, we estimated the models exclusively with the municipalities with foreign companies operating during the 2010–2016 period, therefore excluding smaller (i.e., less populated/developed) regions. Moreover, we excluded the city of São Paulo and the São Paulo metropolitan region, as these municipalities represent significantly well-developed and populated regions. São Paulo city and its neighborhood are responsible for a significant share of the state's GDP and CO<sub>2</sub> emissions and therefore represent dissident regions compared to the rest of the state. In sum, results from Table 4 confirm the effects of GDP and FDI on the environment and are aligned with Table 3.

Table 4. DK estimations for the sub-samples.

Variables	Municipalities with FDI during the 2010–16 Period			Excluding São Paulo City			Excluding São Paulo Metropolitan Region		
	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
GDPPC	0.182 *** (0.0477)	1.714 *** (0.125)	0.180 *** (0.0465)	0.182 *** (0.0477)	1.714 *** (0.125)	0.179 *** (0.0464)	0.210 *** (0.0524)	2.048 *** (0.280)	0.207 *** (0.0506)
GDPPC <sup>2</sup>		−0.214 *** (0.0229)			−0.214 *** (0.0230)			−0.254 *** (0.0418)	
FDI	−0.120 *** (0.0324)	−0.123 *** (0.0305)	0.0884 (0.0841)	−0.120 *** (0.0324)	−0.123 *** (0.0305)	0.0886 (0.0844)	−0.120 *** (0.0324)	−0.125 *** (0.0310)	0.102 (0.0892)
FDI <sup>2</sup>			−0.211 *** (0.0740)			−0.211 *** (0.0742)			−0.221 *** (0.0841)
IND	0.00849 (0.0342)	−0.00715 (0.0420)	0.00931 (0.0329)	0.00822 (0.0348)	−0.00755 (0.0427)	0.00907 (0.0335)	0.0127 (0.0384)	0.000710 (0.0474)	0.0149 (0.0377)
SERV	0.234 *** (0.0616)	0.218 *** (0.0478)	0.228 *** (0.0632)	0.234 *** (0.0617)	0.218 *** (0.0484)	0.229 *** (0.0633)	0.309 *** (0.0772)	0.307 *** (0.0471)	0.302 *** (0.0791)
RES	−0.0748 *** (0.0148)	−0.0720 *** (0.0145)	−0.0752 *** (0.0145)	−0.0744 *** (0.0149)	−0.0717 *** (0.0145)	−0.0748 *** (0.0146)	−0.0669 *** (0.0171)	−0.0648 *** (0.0170)	−0.0667 *** (0.0169)
DENS	0.726 (0.657)	0.782 (0.605)	0.726 (0.654)	0.722 (0.657)	0.779 (0.605)	0.722 (0.653)	0.565 (0.682)	0.623 (0.613)	0.560 (0.676)
Constant	−4.993 (3.939)	−7.883 ** (3.741)	−4.985 (3.910)	−4.951 (3.928)	−7.840 ** (3.729)	−4.942 (3.898)	−4.074 (3.684)	−7.569 ** (3.703)	−4.041 (3.647)
F	51.36 ***	1189.08 ***	14.29 ***	50.86 ***	1189.18 ***	55.28 ***	23.90 ***	1329.63 ***	10.30 ***
Hausman	63.40 ***	37.74 ***	78.87 ***	62.50 ***	32.85 ***	77.97 ***	43.59 ***	93.89 ***	61.90 ***
Mod. Wald.	4516 ***	18185 ***	4584 ***	4503 ***	18120 ***	4576 ***	4067 ***	9411 ***	4344 ***
Wooldridge	58.98 ***	47.52 ***	58.76 ***	58.67 ***	47.29 ***	58.45 ***	56.56 ***	43.60 ***	56.77 ***
Pesaran	128.97 ***	120.66 ***	128.38 ***	127.32 ***	119.04 ***	126.74 ***	94.41 ***	87.71 ***	93.73 ***
Observations	1001	1001	1001	994	994	994	784	784	784
Municipalities	143	143	143	142	142	142	112	112	112

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Standard errors in parentheses. DK: Fixed-effects regression with Driscoll-Kraay standard errors. Sub-sample 1: only municipalities with FDI during the whole time period. Sub-sample 2: São Paulo city excluded. Sub-sample 3: São Paulo metropolitan region excluded.

The results from Table 4 confirm the findings in Table 3. In sum, GDP seems to present a non-linear relationship with CO<sub>2</sub>. An inverted U-shaped relationship between GDP and CO<sub>2</sub> seems to represent the Brazilian reality similar to the original Simon Kuznets [76] curve. Moreover, FDI is, again, linear and negatively associated with emissions, thus being environmentally beneficial. Once more, before declaring support for the hypotheses, we perform robustness checks.

### 5. Robustness Checks

The robustness checking section is twofold. First, we verify if endogeneity is an issue. Second, we evaluate if a non-linear relationship is significant using the threshold regression technique, as some scholars presented criticism towards the use of quadratic variables to test for non-linearity.

The main explanatory variables (GDP per capita and FDI) were tested for endogeneity. Sustainability studies showed that economic growth and environmental degradation may suffer from simultaneity problems. Additionally, the international business literature argues that foreign companies do not choose their subsidiaries' locations randomly. Foreign firms may be attracted to some of the host's characteristics [26]. If foreign firms are choosing to invest in more developed regions, and these regions generate more pollution, then the model may suffer from reverse causality.

Therefore, the C-statistic test (also known as the difference-in-Sargan test) was employed to verify endogeneity issues [77] and we employed the lagged values as instruments, following Yue et al. [78] and Song [79]. Previous scholars argue that finding instruments to employ in CO<sub>2</sub>-related studies is difficult, but defend that past outputs should not impact present emissions. Therefore, one-year lagged values of GDP per capita and FDI were used. Results (see Appendix B) show that FDI is exogenous, but endogeneity is significant for GDP per capita.

To mitigate this issue, we employed the Limited Information Maximum Likelihood (LIML) estimation technique. Previous studies commented that the LIML technique is better than the traditional Two-Stages Least Squares (2SLS) technique, and more capable of ensuring consistency and asymptotic normality [80]. LIML presents good results even in smaller samples [81] and reduces the bias of the estimations [82].

The results presented in Table 5 demonstrate that our linear results are consistent, as they are aligned with the previous estimations in Table 3. GDP per capita is positively associated with CO<sub>2</sub> emissions. Conversely, FDI is negatively associated with emissions.

Furthermore, there is the issue of possible non-linear relationships between our explanatory variables and CO<sub>2</sub> emissions. Previously, we employed the quadratic forms of GDP and FDI in the models, as it is done by the majority of EKC studies [7,15,62]. Recently, however, scholars have commented on the limitations of this method. Girma [83] argues that, although the quadratic specification tends to be more informative than the linear model, it establishes a very specific functional form on the non-linearity that may not be a good representation of the pattern in the data. He and Yao [16] pointed out that, although EKC is traditionally measured in quadratic or even cubic forms, this specification is arbitrary and subjective. For these reasons, scholars are now using more novel methods such as nonparametric regressions and the Non-linear Autoregressive Distributed Lag (NARDL) in non-linear studies [84]. Particularly, the threshold regression is being employed by CO<sub>2</sub>-related studies [16,85,86].

Thus, this study also investigates the non-linear behavior of the variables using the fixed-effect panel threshold regression proposed by Hansen [87] as presented in Equations (6) and (7).

$$\ln CO_{2jt} = \beta_0 + \beta_1 \ln GDP_{PCjt} I(GDP_{PCjt} \leq \lambda) + \beta_2 \ln GDP_{PCjt} I(GDP_{PCjt} > \lambda) + \beta_3 FDI_{jt} + \beta_4 \ln' X_{jt} + a_j + e_{jt} \quad (6)$$

And,

$$\ln CO_{2jt} = \beta_0 + \beta_1 \ln FDI_{jt} I(FDI_{jt} \leq \lambda) + \beta_2 \ln FDI_{jt} I(FDI_{jt} > \lambda) + \beta_3 GDP_{PCjt} + \beta_4 \ln' X_{jt} + a_j + e_{jt} \quad (7)$$

where  $I(\cdot)$  is an indicator function. We follow Girma [83] and employ a set of quantiles (1%, 1.25%, 1.50%, . . . , 98.75%; 99%) to estimate the threshold parameter ( $\lambda$ ). Then, we test the null hypothesis of no non-linearity ( $H_0: \beta_1 = \beta_2$ ) using the likelihood ratio test statistics and their bootstrapped  $p$ -values.

Table 6 shows that a non-linear relationship is only significant for the GDP–CO<sub>2</sub> relationship, with one significant threshold point.

Table 5. LIML estimations.

Variables	Whole Sample		Municipalities with FDI during the 2010–16 Period				Excluding São Paulo City			Excluding São Paulo Metropolitan Region		
	(16)	(17)	(18)	(19)	(20)	(21)	(22)	(23)	(24)	(25)	(26)	(27)
GDPPC <sup>†</sup>	0.526 *** (0.105)	1.356 ** (0.592)	0.527 *** (0.106)	0.696 ** (0.289)	0.649 (0.552)	0.696 ** (0.290)	0.695 ** (0.290)	0.649 (0.552)	0.696 ** (0.290)	1.002 ** (0.504)	0.785 (0.596)	1.002 ** (0.505)
GDPPC <sup>2†</sup>		−0.127 (0.115)			0.00735 (0.118)			0.00733 (0.118)			0.0253 (0.106)	
FDI	−0.0989 ** (0.0451)	−0.130 ** (0.0567)	−0.169 (0.153)	−0.139 * (0.0783)	−0.136 * (0.0821)	−0.0115 (0.207)	−0.139 * (0.0782)	−0.135 * (0.0821)	−0.0106 (0.207)	−0.168 * (0.0899)	−0.151 * (0.0841)	−0.118 (0.244)
FDI <sup>2</sup>			0.0712 (0.144)			−0.128 (0.183)			−0.128 (0.183)			−0.0485 (0.207)
IND	−0.00756 (0.0240)	−0.0952 (0.0826)	−0.00779 (0.0241)	0.0155 (0.0746)	0.0128 (0.0680)	0.0170 (0.0748)	0.0154 (0.0749)	0.0126 (0.0683)	0.0168 (0.0751)	0.0517 (0.106)	0.0342 (0.0811)	0.0524 (0.106)
SERV	0.551 *** (0.0796)	0.980 *** (0.320)	0.552 *** (0.0799)	0.454 * (0.255)	0.433 (0.304)	0.452 * (0.256)	0.454 * (0.256)	0.433 (0.305)	0.453 * (0.257)	0.759 * (0.404)	0.636 (0.406)	0.758 * (0.406)
RES	−0.237 *** (0.0706)	−0.254 *** (0.0773)	−0.237 *** (0.0706)	−0.0940 *** (0.0256)	−0.0928 *** (0.0280)	−0.0939 *** (0.0257)	−0.0934 *** (0.0256)	−0.0922 *** (0.0280)	−0.0933 *** (0.0256)	−0.0952 *** (0.0277)	−0.0893 *** (0.0277)	−0.0951 *** (0.0278)
DENS	0.450 *** (0.164)	0.0114 (0.416)	0.452 *** (0.165)	0.246 (0.339)	0.245 (0.327)	0.245 (0.340)	0.240 (0.340)	0.239 (0.327)	0.239 (0.340)	−0.0767 (0.453)	−0.0434 (0.407)	−0.0781 (0.453)
F	17.59 ***	10.56 ***	15.17 ***	3.19 ***	4.28 ***	2.98 ***	3.17 ***	4.26 ***	2.97 ***	2.24 **	3.17 ***	2.09 **
Observations	3552	3552	3552	858	858	858	852	852	852	672	672	672
Municipalities	592	592	592	143	143	143	142	142	142	112	112	112

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Standard errors in parentheses. † Predicted using one-year lagged values as instruments.

**Table 6.** Test results for the threshold effects.

Variables	Threshold	RSS	MSE	F	p-Value
GDPPC	Single ( $\lambda_1$ )	154.769	0.037	20.300	0.077
	Double ( $\lambda_2$ )	154.374	0.037	10.580	0.340
	Triple ( $\lambda_3$ )	154.037	0.037	9.050	0.573
GDPPC <sup>†</sup>	Single ( $\delta_1$ )	102.709	0.029	24.930	0.054
	Double ( $\delta_2$ )	102.410	0.028	10.350	0.636
	Triple ( $\delta_3$ )	102.186	0.028	7.780	0.858
FDI	Single ( $\phi_1$ )	155.377	0.038	4.040	0.773
	Double ( $\phi_2$ )	155.250	0.038	3.370	0.760
	Triple ( $\phi_3$ )	155.178	0.038	1.920	0.937

<sup>†</sup> Predicted using one-year lagged values as instruments.

Therefore, the threshold tests confirm the linear relationship between FDI and CO<sub>2</sub>, and FDI seems to be negatively associated with emissions. Therefore, the results support H1a and reject H1b.

Table 7 shows the estimated coefficients for GDP following two distinct regimes: lower and higher levels of GDP. These results suggest that, indeed, there is a non-linear relationship between GDP and CO<sub>2</sub>. After the threshold point, the impact of GDP on the environment is lower, meaning that richer regions (i.e., more developed) tend to have a smaller environmental impact, which is beneficial to the region.

**Table 7.** Panel threshold estimations.

Variables	(28)	(29)
GDPPC ( $\leq \lambda_1$ )	0.351 *** (0.0543)	
GDPPC ( $> \lambda_1$ )	0.307 *** (0.0508)	
GDPPC <sup>†</sup> ( $\leq \delta_1$ )		0.636 *** (0.117)
GDPPC <sup>†</sup> ( $> \delta_1$ )		0.574 *** (0.108)
FDI	−0.0585 (0.0792)	−0.0874 (0.0581)
IND	0.0456 * (0.0247)	0.0536 ** (0.0209)
SERV	0.493 *** (0.0655)	0.279 *** (0.0483)
RES	−0.218 ** (0.0952)	−0.222 ** (0.104)
DENS	1.022 *** (0.211)	0.753 *** (0.197)
Constant	−6.018 *** (0.711)	−4.925 *** (0.718)
F	23.47 ***	13.33 ***
Observations	4144	3552
Municipalities	592	592

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Robust standard errors in parentheses. <sup>†</sup> Predicted using one-year lagged values as instruments.

Nonetheless, the threshold regression does not support the EKC hypothesis, as GDP still presents positive effects in both regimes. Hence, Table 7 shows support for H2 and rejects H3. São Paulo state does not present the inverted-U curve that European [88] and Asian [5,7,62] regions present. In sum, the GDP–CO<sub>2</sub> relationship is not linear or quadratic. The policy implications of these findings will be discussed in the next section.



## 6. Conclusions and Policy Implications

The findings show that the relationship between economic growth and emissions is environmentally costly and non-linear. To some extent, GDP increases CO<sub>2</sub> emissions, which corroborates with previous studies' understanding [21,89]. Other studies proposed an inverted U-shape curve between economic growth and emissions [5,8,15,19]. Our study contributes by showing that GDP is non-linear with CO<sub>2</sub> emissions, although this relationship did not present the U-shape form previously proposed. In this sense, we found that GDP becomes less environmentally damaging after a certain threshold point. This result supports hypothesis 2 (*H2: Economic growth presents a non-linear relationship with CO<sub>2</sub> emissions*).

Additionally, the results reveal a negative impact of FDI on CO<sub>2</sub> emissions, therefore supporting hypothesis 1 (*H1a: FDI is negatively associated with CO<sub>2</sub> emissions*). This result corroborates with previous studies that advocate in favor of FDI to reduce pollutant emissions [6,21,23,78]. In contrast, our main finding presents evidence against studies arguing that foreign investments increase environmental damage [7,8,16,19,62,90]. The dataset we employed shows that the majority of foreign companies operating in São Paulo state have origins in developed economies (e.g., United States, Germany, and Japan), and are operating in medium and high-tech sectors (e.g., cars, energy, mobile technologies, and electronics). Therefore, the negative impact of FDI on CO<sub>2</sub> emissions might result from multinational companies with green practices and higher levels of productivity. This is relevant because public authorities in developing regions must create specific policies to attract international enterprises to generate green growth conditions. We argue that FDI must be one of the major drivers to promote economic growth along with sustainable development.

There are two main useful contributions for policymakers. First, the non-linear effect of GDP on emissions might encourage public authorities to find alternative economic growth strategies. Arguably, local policymakers have an essential role in boosting economic growth while reducing power consumption, as energy policies should be managed at the local level [91]. Second, FDI is an essential policy tool to create green development paths in Sao Paulo state. In this sense, local policies must focus on strategies to attract green technologies [92], clean energy sources [93], green jobs and products [94], and higher productivity [95]. Note that FDI is crucial to promote economic growth because it guarantees access to foreign firms' knowledge and technologies to transform the economic structure. For example, the industry is a relevant sector to the Sao Paulo region. Public authorities could reduce taxes to national and multinational companies that emit less CO<sub>2</sub>. A tax policy could also lead green companies to regions that are already presenting higher levels of CO<sub>2</sub>, such as the Sao Paulo city and densely populated regions [96]. New firms might also create local new jobs, increase the local income, and generate green economic growth paths. This new cycle would generate development with lower CO<sub>2</sub> emissions. Besides, alternative transport policies are required in metropolitan areas, such as efficient public transport for passengers, vehicle utilization improvements, and fuel efficiency to reduce pollutant emissions [75,97]. The public sector might invest in environmental projects to develop national green products, such as photovoltaic cells, smart grids, bioplastics, among others [98]. Also, the vast regions in Sao Paulo countryside could focus on alternative energy resources [99], such as the electric energy from sugarcane mills or by building photovoltaic power plants.

Noteworthy, Sao Paulo is the most developed state in Brazil. It means that Sao Paulo has adequate infrastructure (roads, railways, ports, and airports) to connect the region with the entire world. Also, Sao Paulo presents one of the highest human development in Brazil, which is crucial to create human capital (education, health, income, among others). These conditions favor the attraction of FDI to Sao Paulo state. Other emerging regions may use the policy recommendations presented here to facilitate their development.

Finally, some limitations of this study are expected to open multiple avenues for future research. First, this study measures CO<sub>2</sub> emissions through consumption. Future studies are encouraged to examine the impact of FDI and GDP on other types of emissions (e.g.,

sulfur dioxide emissions (SO<sub>2</sub>), air quality, or ecological footprint). Second, this study did not verify how renewable and non-renewable energy sources can individually affect São Paulo state's emissions. Third, we did not control the effect of economic structure and the type of FDI on pollution.

Despite these limitations, this study contributes to the literature on growth, foreign investments, and CO<sub>2</sub> emissions. As previously described, there is no consensus on the effects of growth and FDI on the environment, particularly in emerging economies. To the best of our knowledge, this is the first study to apply Brazilian regional FDI data to better comprehend its environmental consequences. To perform such an analysis, a handmade and unique dataset was built, and more novel econometric techniques were employed. Our findings have important policy implications and are arguably capable of guiding future research in developing regions.

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## Appendix A

**Table A1.** Variance Inflation Factor (VIF).

Variables	Whole Sample		Municipalities with FDI during the 2010–16 Period		Excluding São Paulo City		Excluding São Paulo Metropolitan Region	
	VIF	1/VIF	VIF	1/VIF	VIF	1/VIF	VIF	1/VIF
GDPPC	1.850	0.541	1.330	0.752	1.330	0.752	1.550	0.645
FDI	1.090	0.917	1.140	0.877	1.140	0.877	1.130	0.885
IND	1.810	0.552	1.750	0.571	1.740	0.575	1.860	0.538
SERV	1.500	0.667	2.050	0.488	2.200	0.455	2.200	0.455
RES	1.390	0.719	1.580	0.633	1.570	0.637	1.530	0.654
DENS	1.720	0.581	1.140	0.877	1.260	0.794	1.400	0.714
Mean	1.560	-	1.520	-	1.510	-	1.610	-

## Appendix B

**Table A2.** Endogeneity tests.

Variable	Sample	C-Statistic	p-Value	Result
GDPPC	Whole sample	12.442	0.0004	Endogenous
	Municipalities with FDI during the 2010–16 period	5.741	0.0166	Endogenous
	Excluding São Paulo city	5.713	0.0168	Endogenous
	Excluding São Paulo metropolitan region	5.917	0.0150	Endogenous
	Whole sample	0.096	0.7571	Exogenous
FDI	Municipalities with FDI during the 2010–16 period	0.008	0.9299	Exogenous
	Excluding São Paulo city	0.008	0.928	Exogenous
	Excluding São Paulo metropolitan region	0.843	0.3584	Exogenous

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