AN ANALYSIS OF ENERGY EFFICIENCY IN MULTICAMPI HIGHER EDUCATION INSTITUTIONS AND A NOVEL ENVIRONMENTAL LABELING PROPOSAL

ABSTRACT
The expansion of Brazilian higher education institutions (HEIs) organized in multicampi structures brought a significant complexity to the academic and administrative management. In this context, environmental management strategies become quite relevant, especially when considering the low effectiveness of the Brazilian Labeling Program (BLP) for the classification of buildings in practice. The main objective of the present paper is to evaluate the BLP efficacy as applied to HEI buildings, aiming to develop a new environmental labeling model for multicampi HEIs. For this purpose, the BLP was applied to the labeling of Instituto Federal de Educação, Ciência e Tecnologia do Piauí (IFPI), employing data obtained from electricity bills between 2016 and 2018. The energy diagnosis was performed considering distinct indicators, performance indexes, levels and rankings of relative energy efficiency, from which energy efficiency labels could be developed. The results allow the verification of the low efficiency of the BLP, especially in the environmental labeling of HEIs. From the detailed analysis of bills, it is possible to develop environmental labels inspired by the BLP, resulting in a different approach. A novel type-II environmental labeling methodology is then introduced based on the breakdown of electricity bills and statistical methods.

Keywords: buildings; energy management; labeling.
INTRODUCTION

In the 2030 Agenda published by the United Nations (UN) in September 2015, member countries were committed to 17 main sustainable development goals (SDGs) and 169 other goals to be pursued in the subsequent 15 years (UN, 2015b). The seventh SDG is defined in terms of accessible and clean energy policies, aiming to provide universal access to energy, increase the share of renewable energies, double the global energy efficiency rate, as well as to strengthen international cooperation in research and technology transfer (UN, 2015a). In the context of electricity, the energy efficiency of residential, commercial, and public buildings becomes relevant, which also comprises higher education institutions (HEIs). This aspect is of major concern, especially because the consumption of electricity in buildings accounted for 42.6% of overall Brazilian consumption in 2018 (EPE, 2019, p. 40). Energy efficiency can then be regarded as a set of measures aiming at the development of activities or product and service supply with low energy consumption (PÉREZ-LOMBARD; ORTIZ; VELÁZQUEZ, 2013).

The Brazilian higher education scenario has faced a significant geographical expansion since the 1960s, thus requiring improved academic and administrative management policies to ensure good performance in terms of a multicampi structures (NEZ, 2016). This expansion has become more expressive in federal HEIs since 2007 with the introduction of the Support Program for Restructuring and Expansion Plans of Brazilian Federal Universities (Programa de Apoio a Planos de Reestruturação e Expansão das Universidades Federais — RE-UNI). This program counted with the participation of all federal institutions since its first year of implementation, which led to a significant increase in the existing infrastructure (LIMA; MACHADO, 2016). Since the use of energy is the input with the highest environmental risk index (ERI) in HEIs (SENNA et al., 2014), it is important to consolidate the environmental management taking into account energy efficiency policies (SILVA et al., 2016). This aspect is expected to allow HEIs to promote significant changes in social reality, not only with its core activity, but also with the proper application of management strategies to the buildings in order to consolidate them as sustainable educational spaces.

According to distinct energy balance reports, e.g., the ones presented by the International Energy Agency (IEA, 2018) at the global level and by the Energy Research Company (EPE, 2019) at the national level, energy statistics are generally compiled and presented on a sectoral basis, where consumers are grouped according to the economic activity, i.e., industry, transportation, services, agriculture, and residential consumers. However, the sector in which HEIs are inserted is the most heterogeneous one. Although energy consumption data for the whole sector are available, specific reports for the subsectors are not easily found. In this context, the energy labeling becomes relevant as an environmental management instrument, which can be adopted as a benchmark strategy and encourage the continuous improvement of energy management in HEIs.

According to the Brazilian Association of Technical Standards (ABNT, 2002), environmental labels are statements that indicate the environmental performance of general-purpose products, or even products related to specific environmental aspects, among which the use of energy is included. Energy labels can be employed in the form of text, symbols, or graphic elements added to packages, technical bulletins, or advertisement material to allow a fair comparison among products of the same type, aiming at promoting conscious consumption. Such labeling can also be classified in three categories: type I, which refers to third-party labeling, according to the fulfillment of specific requirements (ABNT, 2004); type II, which corresponds to environmental self-declarations presented by product and service suppliers without independent certification (ABNT, 2017); and type III, which refers to the labeling of products or services while presenting environmental information on the life cycle, also comprising several stages from the acquisition of raw material to their final disposal (ABNT, 2015).

The first milestone in environmental labeling was established in 1977, when Germany created the Blue Angel label. Since then, similar strategies have been adopted in several countries worldwide, e.g., the Green Label in the United States of America (USA), the EcoMark label established in Japan in 1989, and the EU Ecotool created by the European Union in 1998 (PRIETO-SANDOVAL et al., 2020). According to Spengler et al. (2020), the Blue Angel label is still in force and can be classified as type I. This label is applied to dis-
tinct products to reduce impacts during the utilization stage, whereas other stages of the product life cycle are covered to a lesser extent.

Figure 1 shows a timeline comprising regulatory frameworks that led to the current state of energy consumption labeling in Brazilian buildings. In order to develop a benchmarking strategy in Brazil, the Brazilian Labeling Program (PBE) was created in 1984. Under the supervision of the National Institute of Metrology, Quality and Technology (Inmetro), which was created in 1973, and the Ministry of Mines and Energy (MME), this program is dedicated to provide consumers with information on energy efficiency of distinct products. Thus, it intends to encourage a more conscious purchasing process through the National Energy Conservation Label (ENCE), which can be classified into two types: voluntary or compulsory certifications (EPE, 2020). The social benefits of voluntary labeling increase proportionally with the market demand for certified products, the relevance of labeling issues, and the consumer ability to understand, trust, and interpret the provided information on the label. On the other hand, compulsory labeling is associated to several negative aspects, e.g., high social and production costs; extremely rigid production patterns; increase in size and complexity of the supply chain; sudden changes in production patterns or techniques; and market distortions with reduced competition (ROE; TEISL; DEANS, 2014).

The BLP is supported by the National Electricity Conservation Program (Programa Nacional de Conservação de Energia Elétrica – Procel), which was created in 1985 with the subprogram named Procel Seal (Selo Procel). This practice was instituted in 1993 to classify distinct equipment according to a given efficiency index (ELETROBRAS, 2019). Therefore, the National Label for Energy Conservation (Etiqueta Nacional de Conservação de Energia – ENCE) is regarded as the certification provided by Inmetro to several products, whose energy efficiency performance can be classified according to proper levels. Procel Seal is then used to identify the best products in a given category according to a type-I environmental label, highlighted at the end of the timeline shown in Figure 1.

In addition to Procel Seal, Procel Energy Efficiency in Buildings (Procel Edifica) subprogram was created in 2003 for the development of activities that promote the energy efficiency improvement in buildings. It is responsible for coordinating the Brazilian Building Labeling Program (BLP Edifica), which aims at the environmental labeling of energy efficiency in Brazilian buildings (EPE, 2020). The creation of the Energy Effi-
ciency Program (EEP) for Electricity Distribution Companies (BRASIL, 2000) and the enactment of the Energy Efficiency Law (BRASIL, 2001) must also be highlighted. As from the introduction of this policy, electric energy distribution companies are expected to employ a minimum percentage of the net operating revenue (NOR) for the development of energy efficiency programs and Procel. Besides that, it also represents an important source of funds for the program. On the other hand, the Energy Efficiency Law establishes the minimum energy efficiency requirements for machines and devices powered by electricity, which are manufactured or marketed in the country, as well as a program of goals for improving this classification (BRASIL, 2001). It is directly associated to the BLP and Procel, given that such mechanisms tend to encourage the acquisition of energy-efficient products.

As a direct consequence of such policies, an energy saving of 21.2 billion kWh was reported in 2017, corresponding to an amount 39.89% greater than that of the previous year. It is also equivalent to 4.57% of the overall electricity consumption in Brazil during 2017, as associated with the annual consumption of 11.25 million households. This reduction accounts for the emission of 1.96 million tons of CO$_2$ into the atmosphere, i.e., 675,000 cars over one year (ELETROBRAS, 2018). However, in order to increase the effectiveness of the energy efficiency policies, Nogueira et al. (2015) underline the need for marketing campaigns to inform the general public about regulation benefits; constant review of efficiency limits; increase in the amount of regulated equipment; and increase in the number of equipment compulsorily withdrawn from the market due to low energy efficiency, similarly to what occurred with incandescent lamps.

The BLP initiatives have contributed to the efficient use of energy in the country. However, this certification can also be applied to buildings in Brazil since 2009 with the ENCE within the scope of the BLP Edifica as a result of the interaction between the BLP and Procel Edifica. The Management Committees for Energy Efficiency Indicators and Levels (Comitê Gestor de Indicadores de Eficiência Energética — CGEE), which were established by the Energy Efficiency Law (BRASIL, 2001), joined the Technical Building Commission (CT-Edificações) together with Inmetro in 2005, considering that it also established the regulatory frameworks of BPL Edifica.

The Technical Quality Requirements for the Energy Efficiency Level of Commercial, Service, and Public Buildings (RTQ-C) were published in 2009 and modified in 2010 (INMETRO, 2010). The Technical Quality Regulation for the Energy Efficiency Level of Residential Buildings (RTQ-R) was also published in 2010. This regulation and its complementary documents were properly modified two years later (INMETRO, 2012).

The aforementioned regulations use engineering calculations and computer simulations from three individual building systems, i.e., envelope, air conditioning, and lighting to determine the energy performance of a part or the whole building during the design or conclusion stage of the facility. It considers a set of mandatory requirements, e.g., insulation of air conditioning system pipes, as well as bonus issues corresponding to photo-voltaic generation systems for supplying at least 10% of the overall energy consumption, for instance.

Borgstein, Lamberts and Hensen (2016) could identify 17 examples of energy performance classification systems and projects in non-residential buildings in five countries (Brazil, the United States, the United Kingdom, China, and Australia), among which BLP Edifica is included. In this study, the program was characterized as a labeling system with asset classifications, which are applicable to new buildings and major renovations based on prescriptive or simulation methods for a comparison with national buildings. With a focus on comparing the energy consumption of buildings, these assessment techniques are used to classify, tax, and rank the energy performance of buildings on a compulsory or voluntary basis. In most cases, it also aims to ensure good performance levels and reduction of energy consumption in buildings, in addition to providing a proper classification according to the type (simple labeling or benchmarking, for example), application, and benchmarking methodology.

There are several methodologies for assessing the energy performance of buildings, which can assume several classifications based on the adopted techniques. According to Hong et al. (2013) and Burman et al. (2014), such strategies can be classified as top-down or bottom-up. In the first approach, the benchmarks are obtained based on the energy performance indices of the building. The second one considers that the indices can be obtained from the theoretical analysis of a building. According to the control level of the de-
termination processes, Li, Han and Xu (2014) state that methodologies can be classified into white, gray, or black box type. In the first model, the techniques are purely statistical, with little information required for each building. In the second model, a limited physical analysis of the building is mixed with statistical methodologies. The last model relies strongly on the physical structure of buildings and is highly dependent on input data provided by the user. The choice of a given methodology must be performed according to the objective of the study, which is limited by the availability of data and cost-benefit ratio of data acquisition.

Among the existing techniques, it is possible to identify and classify some of the most common methodologies for determining the energy performance of non-residential buildings: engineering calculations (bottom-up and white box) (Fucci et al., 2016); statistical methods (top-down and gray or black box) (Li; Tao, 2017); simulations (bottom-up and gray or black box) (Planas; Cuerva; Alavé-Dra, 2018); machine learning (top-down and black box) (Cheng et al., 2019); dynamic methods and real-time analysis (top-down and black box) (Schibuola; Scarpa; Tambani, 2018); load curve analysis and energy bill breakdown (top-down and black box) (Niederberger; Champniss, 2018); energy audit (bottom-up and white box) (Mathioulakis et al., 2017); post-occupation, comfort, and environmental quality analysis (bottom-up and gray or black box) (GhaHRamani et al., 2018). According to the purpose of the study, the methodologies can be independent or complementary.

Public assessment systems for energy performance can be used in new and existing buildings for distinct purposes, which include mandatory compliance with minimum requirements; conception of policies to penalize low energy performance, or reward good energy performance; mandatory or voluntary labeling of the energy performance of buildings; transparency and disclosure programs; voluntary internal benchmarking; and evaluation of improvement opportunities (Borgstein; Lamberts; Hensen, 2016). BLP Edifica is characterized as one of the systems, which is an important instrument to ensure the good energy efficiency performance of buildings in Brazil. However, there is the need for a simplified, more accurate and efficient model for the labeling process (MeLo et al., 2012). Thus, for achieving the full implementation of this program, there is a need for developing methodologies to calculate realistic values; creating awareness campaigns for the general public; providing training and support to increase the number of energy assessors; applying sanctions in case of non-compliance with the existing regulation; conducting post-certification monitoring and evaluation measures; establishing and maintaining a central recording system for collecting relevant certification data (Wong; Krüger, 2017).

From the comparison of the Brazilian labeling model with those used by other countries, e.g., the United States and Portugal, it is clear that motivating the conception of zero energy consumption buildings, which are totally self-sufficient; informing CO₂ emission reductions; expanding its coverage to the industrial sector; and making certification mandatory is required (Lopes et al., 2016). It is also worth mentioning that labeling buildings is still voluntary in Brazil. However, compliance with ENCE became mandatory since June 2014 for purchasing or renting electric equipment and devices, as well as for new building projects for federal institutions, which are facilities subject to retrofit with federal public resources for the exercise of administrative activity, or for the provision of public services (Brasil, 2014). Therefore, federal public facilities can be regarded as any property built or adapted with federal public resources for the exercise of administrative activity or for the provision of public services. Even with such mandatory requirements, BLP Edifica is ineffective in labeling the energy efficiency of buildings.

Six years after the program’s creation, considering a period until 2015, during which the expansion of federal institutes of education, science and technology occurred, only 84 commercial, public, and service buildings were labeled throughout the country in a few states, which do not include Piauí State, as stated by Silva et al. (2015) and Wong and Krüger (2017). In this context, the present study aims to provide answers to the following question: How is it possible to develop an energy efficiency label for compliance with environmental management in multicampi HEIs in a reliable and effective manner? The main hypothesis is that energy diagnosis may rely on the detailed analysis of electricity bills, as well as productivity indicators for the conception of energy efficiency indices and levels that can be summarized in environmental labels inspired by the BLP, which is a different methodology than that of standardized by Inmetro (2010). Therefore, a type-II
A new environmental labeling methodology is introduced in this paper, based on the breakdown of electricity bills, as well as top-down and gray box statistical methods.

For this purpose, the general objective of this study consists in evaluating the effectiveness of the BLP applied to the energy efficiency labeling of HEI buildings for developing a new environmental labeling model especially dedicated to multicampi HEIs. IFPI, which is held by the Brazilian Federal Government, was chosen as the analyzed scenario (Silva et al., 2018).

**MATERIALS AND METHOD**

Research started with the assessment of the BPL applied to HEIs. An extensive search was performed in Inmetro website (INMETRO, 2015) to identify and classify the ENCE regulations issued for buildings in Brazil, specifically comprising HEIs. Then, a general characterization of IFPI was obtained based on the available bibliography and the institution website. Relevant documents such as the institutional development plan and management reports were used for the energy diagnosis.

According to the methodology used by Silva et al. (2016; 2018) and Silva et al. (2017), a census survey was carried out and, considering the multicampi organizational structure, an extensive analysis of electricity bills of all campuses was developed for the period between 2016 and 2018. Created by the Brazilian Federal Government in 2008, in agreement with the Federal Law No. 11,892, IFPI (2014) aims to “be consolidated as a center of excellence in professional, scientific, and technological education, ranked among the best educational institutions in the Northeastern region”. It currently encompasses 17 campi and 19 energy consuming units, which are distributed over all regions of Piauí State, as shown in Figure 2. All campi are the subject of research, except

![Figure 2 – Description of IFPI campi.](source: adapted from Silva et al. (2018))
for advanced campi and the main administration building, whereas the analysis considers indicators measured from 2016 to 2018.

With the breakdown of 684 electricity bills, a preliminary energy diagnosis was performed while identifying three relevant aspects: manageable costs; avoidable loss and other issues; and proposal of strategies for the energy efficiency evaluation of HEIs. The breakdown of electricity bills is typically used in similar studies, which include those developed by Ketchman et al. (2018) and Niederberger and Champniss (2018). Partial results were then presented to the administration board of IPFI to establish a continuous flow of information in the subsequent months and enable the continuity of studies. In addition, based on the percent participation of cost centers in the total electricity bill, the indicator of avoidable loss (IndAL) was defined as the ratio between manageable costs and the sum of costs associated to avoidable loss.

In order to establish the energy efficiency labels for the campi, an environmental indicator matrix was compiled using the pressure-state-response (PSR) framework (OECD, 2003) based on the pressure (energy use) and state (number of civil servants, infrastructure, and educational aspects) variables, just like recommended by Silva et al. (2018). Such variables were chosen based on the end activity of the institution and on the following qualitative and quantitative criteria: data reliability, relation with the addressed problem, usefulness to the user, data availability, relevance, redundancy, and measurability. Primary data associated with the indicators were used in all campi between 2016 and 2018 in a formal request forwarded to the administration board, in addition to secondary data collected from Nilo Peçanha Platform. This virtual environment is available to the IFPI community since 2017, containing official statistics from the Brazilian Federal Network for Professional, Scientific and Technological Education, which also comprises IFPI (BRASIL, 2019b).

From such data used for classifying the campi in different scale categories, statistical software SPSS was employed to perform a multivariate analysis; and a dendrogram was plotted, which is a graph obtained with the cluster analysis to aggregate different observations. The Euclidean distance was used as an evaluation criterion and the centroid of each cluster of observations was considered. The value of each observation corresponds to the arithmetic mean of the values for the indicator considering 2017 as a reference for all campi (MANLY, 2008). Besides that, in order to prevent a given variable from dominating the analysis, the data were previously normalized based on the difference between the minimum and maximum values (KILKS, 2015).

According to the recommendation provided by Silva et al. (2018), the data envelopment analysis (DEA) (CHARNES; COOPER; RHODES, 1978) was employed with the analysis model (Charnes, Cooper and Rhodes — CCR or Banker, Charnes, and Cooper — BCC), which is based on the homogeneity of samples (CHARNES; COOPER; RHODES, 1978; BANKER; CHARNES; COOPER, 1984). Thus, new efficiency indices were determined: the general indicator of manageable costs (IndMC), which was obtained with the DEA using the BCC model; and the clustering IndMC, which resulted from the DEA based on the CCR model. In addition to this, given the number of observations, the DEA golden rule, as defined by Banker et al. (1989), was applied to the third version of the Integrated Decision Support System (SIAD) (ANGULO MEZA et al., 2005) software. This methodology was used by Blum and Okwelum (2018), Borgstein and Lamberts (2014), and Domingos et al. (2018) in the energy efficiency evaluation based on DEA, aiming at the analysis of household appliances, buildings, and public policies, respectively.

From the aforementioned indicators, the general relative energy efficiency index (REEI) was created and clustered using the arithmetic mean of IndAL and IndMC. By using such new data, the campi were classified according to five levels of relative energy efficiency with equal lengths, based on the difference between the highest and lowest indicators. A general ranking was then established in terms of energy performance while estimating the benchmarks for increasing energy efficiency. The main goal was defined as the elimination of avoidable loss. The DEA models were then used to provide inputs and outputs for establishing energy consumption and demand benchmarks, and for teaching activities, as well as the best efficiency energy level, as observed in Figure 3.

According to the diagram shown in Figure 3, a new type-II environmental labeling model is proposed for energy efficiency evaluation, i.e., a self-declaration approach that is supposed to be employed...
by the HEIs without independent certification while following general and specific requirements provided by ABNT (2017). For this purpose, the model standardized by BLP Edifica, i.e., the traditional paradigm, was used as a starting point. A new energy efficiency label was presented for each of the campus containing relevant information for the energy efficiency analysis, such as indicators, indices, levels, rankings, and diagnostics aiming at a benchmarking strategy.

RESULTS AND DISCUSSION

From data provided by Inmetro (2015), it was possible to identify only 96 non-commercial buildings with the energy efficiency labeling in the scope of the BLP until June 10th, 2019, corresponding to 18 additional units when compared to the amount reported by Wong and Krüger (2017). The Southeastern and Northeastern regions stand out with the highest percentage of issued labels (50 and 24%, respectively), mainly in São Paulo and Bahia States, which were the states with the highest number of ENCE certifications, i.e., 27.1 and 20.8%, respectively. A total of 12 other states with no energy efficiency certification for non-residential buildings were also found, i.e., Acre, Alagoas, Amapá, Goiás, Maranhão, Paraíba, Pernambuco, Piauí, Rio Grande do Norte, Rondônia, Roraima, and Tocantins. From the percent analysis of each energy efficiency level applied to the buildings according to BLP Edifica, it was possible to verify that most of the systems have good performance. Moreover, the percentage of the assessed buildings ranged from 66 to 100%, as observed in Figure 4. This variation is due to the possibility of providing partial certifications to the buildings. However, the envelope analysis was carried out in all buildings considering that this is a mandatory requirement for obtaining general or partial ENCE labels.

Only two HEI buildings received the ENCE label: the School of Architecture of Universidade Federal de Minas Gerais (UFMG) in November 2013; and the University Restaurant of Universidade Federal de Santa Catarina (UFSC) in October 2012. The building envelope and lighting at UFMG were classified with energy efficiency levels C and B, respectively, which are considered low. On the other hand, the building envelope at UFSC was classified with energy efficiency level A, the highest rank.

Figure 3 – Environmental energy efficiency labeling of multicampi HEIs.
It is important to highlight that, even if more than one system is certified, only one ENCE certificate is issued. This result proves that even with the requirement of ENCE in federal HEIs (BRASIL, 2014), the BLP Edifica is quite ineffective in the certification of buildings as evidenced by Silva et al. (2015) and Wong and Krüger (2017), mainly in the case of HEIs, which should be an example of sustainability to the entire community.

### Table 1 – Electricity costs of IFPI between 2016 and 2018.

<table>
<thead>
<tr>
<th>INPUTS</th>
<th>2016</th>
<th>2017</th>
<th>2018</th>
<th>Annual Average</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total bill</td>
<td>R$ 4,330,125.85</td>
<td>R$ 4,624,818.04</td>
<td>R$ 5,921,169.26</td>
<td>R$ 4,958,704.38</td>
<td>100.00</td>
</tr>
<tr>
<td>Total manageable costs</td>
<td>R$ 4,360,863.56</td>
<td>R$ 4,762,192.47</td>
<td>R$ 6,059,927.94</td>
<td>R$ 5,060,994.66</td>
<td>102.06</td>
</tr>
<tr>
<td>Demand</td>
<td>R$ 708,753.53</td>
<td>R$ 648,530.82</td>
<td>R$ 779,652.78</td>
<td>R$ 712,312.38</td>
<td>14.36</td>
</tr>
<tr>
<td>Peak consumption</td>
<td>R$ 1,279,736.25</td>
<td>R$ 1,650,506.04</td>
<td>R$ 2,099,384.28</td>
<td>R$ 1,676,542.19</td>
<td>33.81</td>
</tr>
<tr>
<td>Off-peak consumption</td>
<td>R$ 2,289,969.22</td>
<td>R$ 2,348,814.02</td>
<td>R$ 3,045,944.65</td>
<td>R$ 2,561,575.96</td>
<td>51.66</td>
</tr>
<tr>
<td>COSIP</td>
<td>R$ 82,404.56</td>
<td>R$ 114,341.58</td>
<td>R$ 134,946.23</td>
<td>R$ 110,564.12</td>
<td>2.23</td>
</tr>
<tr>
<td>Total avoidable losses</td>
<td>R$ 266,582.22</td>
<td>R$ 170,546.17</td>
<td>R$ 259,057.56</td>
<td>R$ 232,061.98</td>
<td>4.69</td>
</tr>
<tr>
<td>Exceeding demand</td>
<td>R$ 121,474.48</td>
<td>R$ 27,761.74</td>
<td>R$ 44,475.81</td>
<td>R$ 64,570.68</td>
<td>1.30</td>
</tr>
<tr>
<td>Exceeding reactive power</td>
<td>R$ 69,234.70</td>
<td>R$ 76,354.14</td>
<td>R$ 102,183.25</td>
<td>R$ 82,590.70</td>
<td>1.67</td>
</tr>
<tr>
<td>Fine</td>
<td>R$ 58,466.57</td>
<td>R$ 54,879.87</td>
<td>R$ 81,991.74</td>
<td>R$ 65,112.73</td>
<td>1.31</td>
</tr>
<tr>
<td>Default interest</td>
<td>R$ 9,951.10</td>
<td>R$ 10,721.37</td>
<td>R$ 15,447.24</td>
<td>R$ 12,039.90</td>
<td>0.24</td>
</tr>
<tr>
<td>Monetary correction</td>
<td>R$ 8,260.70</td>
<td>R$ 1,325.42</td>
<td>R$ 15,477.63</td>
<td>R$ 8,354.58</td>
<td>0.17</td>
</tr>
<tr>
<td>Other costs*</td>
<td>-R$ 297,319.94</td>
<td>-R$ 307,920.60</td>
<td>-R$ 397,816.24</td>
<td>-R$ 334,352.26</td>
<td>-6.74</td>
</tr>
</tbody>
</table>

*Additional costs and eventual invoice discounts were not analyzed in the project.

Source: prepared by the author employing data from electricity bills between 2016 and 2018.
However, this study does not take into account the implications of the Energy Efficiency Law (BRASIL, 2001) and the mandatory purchase of products with energy efficiency level A (BRASIL, 2014), which may be the scope of future work.

Considering the multicampi organizational structure of IFPI, electricity costs were estimated based on the breakdown of electricity bills of all campi in 2016, 2017, and 2018, according to Table 1. An annual average value of R$ 232,668.59 was determined, i.e., 4.68% of the total cost, which is higher than the amount spent individually on electricity by six of 17 campi, as shown in Figure 2 (CACAM, CACOC, CAPAU, CAPEDI, CASJP, and CAVAL). In this context, such loss should be promptly eliminated. The amount transferred to the local utility in the form of fines due to delay in the payment of bills is equal to 37%; 35% corresponds to excessive consumption of reactive power, i.e., the amount of energy supplied by the utility that does not result in useful work; and 28% is due to the higher energy consumption than the contracted demand.

Additionally, the annual average manageable costs of R$ 5,060,994.66 must be reduced with the proper execution of energy efficiency projects (EEPs), also comprising environmental education strategies aiming to reduce energy consumption during the more expensive period, i.e., the peak hours between 5:30 p.m. and 8:30 p.m. (ANEEL, 2010). During this interval of only three hours, the amount spent on energy consumption corresponded to 33% of the overall cost. During the three-year period under analysis, there has been a growing trend involving all components associated to manageable costs, as well as in all other components related to avoidable loss, except for the exceeding demand. This component faced a drastic decrease from 2016 to 2017, but it increased again from 2017 to 2018 according to Figure 5A, thus reflecting a specific action taken by the HEI administration board.

The preliminary analysis of energy supply contracts suggests that there was a reduction in the amount paid on the exceeding demand from 2016 to 2017, as a result of the review of the contract terms of all consumer units at the beginning of 2017. However, the increase in this cost in the subsequent year suggests that even though it is a concrete and efficient action, this was a one-off action that should be maintained to ensure good results. The increase in electricity costs also reflects the annual tariff adjustment, which varies significantly. Thus, the energy management, especially in terms of energy efficiency, must consider the variation in consumption and demand of the facilities. However, the analysis in monetary terms is relevant due to the multidisciplinary nature of the study, which is a specific aspect of environmental sciences and involves the impact of costs on the management of organizations.

Breaking down electricity bills was also essential to calculate IndAL, which measures the percentage of avoidable loss in the amounts paid to the utility and corresponds to half of the weight of the REEI as pro-

Figure 5 – (A) Profiles of avoidable loss in all IFPI campi and (B) avoidable loss per campus between 2016 and 2018.

Source: prepared by the author employing data from electricity bills between 2016 and 2018.
posed in the present study. The variation of IndAL in the IFPI campi from 2016 to 2018 is also shown in Figure 5B, in which the green and red colors are used to represent units with improved and worsened performances from 2016 to 2018. It seems that even with the significant reduction of loss associated to the exceeding demand, nine out of 17 campi presented the worst performance in terms of such indicator, thus suggesting the need for an improvement in the energy management strategies of the HEI, especially with respect to electricity costs.

All the actions proposed to eliminate avoidable loss can provide average monthly savings of R$ 19,338.49. They should be considered in an economic analysis as prominent advantages of the proposed interventions, considering important aspects such as the on-time payment of invoices, installation of capacitor banks, and review of the electric energy supply contracts with the local utility. However, the manageable costs represent the largest portion of the electric energy expenses, i.e., almost 21 times higher than the cost of avoidable loss. This aspect requires the implementation of an efficient energy management system. For this purpose, data for the elaboration of an indicator matrix, as defined by Silva et al. (2018) between 2016 and 2018, were used. From the previously normalized state indicators of 2017, a clustering analysis was performed, and two clusters could be formed based on the multivariate distance of the campi, thus denoting that they are all similar to each other, except for CATCE, as shown in Figure 2.

Thus, based on productivity concepts, it was possible to determine IndMC using the measurements related to energy use (inputs) and education (outputs) in 2016, 2017, and 2018, according to Silva et al. (2018). Such data represent a measure of the demand relative efficiency and use of active power, which is associated to the energy used to perform useful work. The DEA BCC model was used to determine the general IndMC, because this model considers different returns depending on the scale of the decision-making unit (DMU). On the other hand, the DEA CCR model was used to determine the clustering IndMC, because it considers constant returns regardless of the DMU scale. The annual variations of this indicator are presented in general terms in Figure 6A and by clusters in Figure 6B, in which the green and red colors denote the campi with worsened and improved performances from 2016 to 2018, respectively. Floriano Campus (CAFLO) could not be included in the analysis because 2016 and 2018, were used. From the previously normalized state indicators of 2017, a clustering analysis was performed, and two clusters could be formed based on the multivariate distance of the campi, thus denoting that they are all similar to each other, except for CATCE, as shown in Figure 2.

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its respective efficiency was not determined in 2016. The utility energy meter installed in the facility could not measure the peak and off-peak energy consumption separately, which are essential variables for calculating IndMC.

It is observed that there is no regular pattern in the variation of IndMC in the general or in the clustering analysis, given that it increased or decreased randomly in some campi. In addition, there are inconsistencies in the comparison between indicators due to the differences between the DEA BBC and DEA CCR models, which were used for determining the general IndMC and clustering IndMC, respectively. DEA BBC considers that the scale returns vary, but DEA CCR provides constant values to such quantities. However, the variations in the indicators related to energy use and education at Angical Campus (CAANG) from 2016 to 2018 can be highlighted, since the general IndMC and clustering IndMC varied -0.23 and -0.24, respectively. Variations of inputs and outputs were also found in Paulistana (CAPAU) and Cocal (CACOC) campi for the same period, with the highest increase in general IndMC and clustering IndMC corresponding to 0.31 and 0.40, respectively.

As previously discussed, the DMU with higher efficiency consequently has the best productivity, being possible to define this concept as the ratio between outputs and inputs. Thus, when analyzing the variation of the IndMC in CAANG, which is the campus with the largest decrease, it is observed that this indicator reflects the variation in productivity. There was an increase in inputs in Figure 7A and a decrease in outputs in Figure 7B. Such aspects cause a decrease in productivity and efficiency. Considering the campi in which IndMC increased, the direct impact on the increased productivity and efficiency is noticeable.

After calculating IndAL, general IndMC, and clustering IndMC, it was possible to determine the REEI in terms of general and clustering values using the arithmetic means of the aforementioned indicators for all campi from 2016 to 2018, except for CAFLO in 2016. In general, an increase in this indicator over the years for both representations can be noted due to the improvement of IndAL. However, the positive impact on this indicator is due to a one-off action, which should be maintained.

Analogously to the indicators that compose it, there was no defined pattern in the variation of the REEIs in IFPI, whereas some units presented improved or worsened performance in the clusters. The results are shown in Figure 8 for all campi in 2018. São Raimundo Nonato campus (CASRN) and CAFLO present the lowest and highest REEIs, respectively, which is also reflected by the existence of a grid-connected photovoltaic system installed in CAFLO (MORAISS et al., 2019).

In order to achieve the maximum energy efficiency level in all campi, the maximum level is required for both IndAL and IndMC. The maximum performance

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**Figure 7** – Behavior of (A) energy indicators – inputs and (B) education indicators – outputs of CAANG between 2016 and 2018.

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*Source: prepared by the author employing data from electricity bills between 2016 and 2018, also considering primary data provided by IFPI (2014) and Brasil (2019b).*
in the first indicator can be obtained only if avoidable loss is completely eliminated, which include important actions, such as the on-time payment of bills, revision of energy supply contracts, eventual installation of capacitor banks, and the consequent elimination of excessive reactive energy consumption. There are two possible solutions to maximize IndMC: reduction in consumption and demand for active energy; or increase in the number of students. Both actions should be developed simultaneously. The goals for decreasing inputs and increasing outputs include the application of DEA to inputs and outputs of campi with low performance in this indicator, as stated by Silva et al. (2018). This action is supposed to bring additional benefits, such as the increase of monetary resources received by the HEI; reduction in the socio-environmental impacts related to the power system expansion; and establishment of a benchmark model for other HEIs in Brazil.

The campi with the lowest IndMC have more ambitious goals toward an efficient energy management strategy. Considering environmental statements can be presented in different ways (ABNT, 2017), relative energy efficiency labels (REELs) were developed as shown in Figure 9. This approach can be regarded as a clear, transparent, sound, and documented type-II environmental labeling proposal, which also comprises some essential characteristics to ensure the self-declaration reliability.

In addition to socioenvironmental benefits, the implementation of energy saving strategies also contributes to mitigating the impacts of contingency of funds. This scenario is currently faced by Brazilian HEIs, which have struggled to reduce operating expenses in the last years (SANTOS, 2020). Thus, the composition of the proposed labels for the multcampi HEIs can be used as an important tool to develop benchmarking strategies and guide the actions toward the conception of an efficient energy management system, in terms of energy efficiency indexes, levels and ranking, as well as energy diagnosis.
Even without considering the implications of the Energy Efficiency Law (BRASIL, 2001) and the mandatory purchase of products with the energy efficiency index class A (BRASIL, 2014), the survey carried out on the Inmetro website has proven the ineffectiveness of the BLP applied to the environmental energy efficiency labeling of buildings in Brazil. This fact is especially true in the case of HEIs, since only two buildings were certified even after ten years of regulation.

The energy diagnosis performed at the IFPI proves that the institution does not adopt an adequate energy management strategy. It was found that 4.68% of the total amount paid to the local electricity company in 2016 to 2018 are due to loss, corresponding to a value higher than that paid for the electric energy consumption in six out of 17 campi in 2018.

The development of research activities in energy efficiency can lead to positive impacts on the energy management in HEIs, which may include review of energy supply contracts, for instance. However, if there is no continuous monitoring of energy consumption, implications may not be fully realized.
in terms of the adoption of an accurate energy management system, sometimes these one-off actions may lead to poor results in the long term.

The breakdown of energy bills can be used in the energy diagnosis of multicampi HEIs, thus allowing the identification of cost centers, which include avoidable loss, manageable costs, among others. From the definition of an indicator of avoidable loss (IndAL), it was possible to determine the campi that have the highest and lowest contributions to the existing expenses. In addition, an environmental indicator matrix using the PSR model could be designed to characterize all campi, making it possible to cluster them with multivariate statistics and to define an indicator of manageable costs (IndMC) using the general and clustering DEA.

Therefore, the matrix data can be used as inputs and products, such as in the form of education and energy use indicators, respectively, aiming at an analysis of productivity and efficient energy use associated with the breakdown of electricity bills. This study resulted in IndAL, which was used to define the general REEI and clustering REEI. From such data, the traditional CCR and BCC DEA models could also be used to determine goals for reducing energy consumption and demand and or increase the education indicators in multicampi HEIs, considering that both actions should be carried out simultaneously.

By using benchmarking, this methodology can be employed to rank the campi and monitor energy consumption in various measurement cycles, thus promoting the continuous improvement of this environmental aspect. In fact, this is the main objective of an energy management system, which may also contribute to the guidelines provided by Agenda 2030. Such results can also be summarized in terms of a REEL, which is classified as a type-II environmental labeling approach applied to each campus. Therefore, this can be regarded as a top-down and black box methodology, which is simpler and more effective than that standardized by Inmetro, i.e., the traditional paradigm. Besides that, the latter approach is sometimes limited given the high complexity, intersubjectivity, and instability of the environment, which confirms the main hypothesis investigated in this study.

Research results are expected to lead to positive impacts on the environmental performance of HEIs. The promotion of awareness of the academic community to environmental issues is also intended, considering that energy consumption is only one of many manageable environmental aspects. Within this context, the development of sustainable spaces in HEIs is highly encouraged. The present study aims to motivate the development of similar studies in other institutions, and the proposed methodology can be used as a benchmark in the energy management of HEIs. Finally, given the flexibility of the introduced techniques, this strategy is expected to be applied to other multisite organizations and aggregate other environmental aspects.

REFERENCES


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