

# Delimitation of Urban Areas with Use of the Platform Google Engine Explorer

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**Abstract:** Google Earth Engine is a cloud computing platform for developing and hosting web applications that allows for the automatic sorting and mapping of terrestrial coverage. The objective of this research is to evaluate the tool's potential for the generation of thematic maps of urban areas, using a big data platform, with data in the cloud. The proposed methodology evaluates the CART classifier for different scales. The local scale considered the area of Rio de Janeiro city. A simplified legend (urban and non-urban) and other with greater detailing (different types of urban intensification), were tested. The main input was the Landsat TOA (Top of Atmosphere) mosaic. Potential, classification time, and results were evaluated. The main products generated were the temporal classifications, in which one can observe the expansion of urban areas and some confusion between classes. In this case editing is necessary. The rapidity in the classification and generation of products is one of the most important positive points of the analysis. The tool is very interactive and easy to handle, even by users with little experience. The urban areas delimitation and identification were promising, requiring more research on the best techniques to be adopted at each geographic scale.

## 1 INTRODUCTION

Geography as a science deals with the analysis of space and the actions of transforming agents of this space.

The traditional mapping based on the use of desktop tools, in actual moment has been complemented by the use of online platforms for data storage and processing.

In this context, cloud computing and the use of Big Data are presented as a potential model for spatial analysis, and mapping.

The Google Earth Engine, and in special the Earth engine Explorer, is a tool with enormous potential to create geographic analyzes, in attending to several applications in the geosciences, and to the Geography.

Among the potential analyzes is the delimitation of urban areas and their expansion, from mappings and the monitoring of these areas.

## 2 RELEVANCE

Given the increasing volume of data generated daily it is important to consider the use of spatial analysis tools that allow integration, ease of access and speed in data processing.

The access and sharing of the web, brings in this way, the discussion about of the Big Data concept of and the online systems for the generation of mappings and geographic analyzes.

The evolution of cloud computing is one of the greatest advances in the history of computing, becoming a new paradigm in recent years. Among the various existing definitions, cloud computing can be defined as a collection of virtualized and interconnected computers that provide computing resources and services and are dynamically provisioned and presented based on an agreement between the service provider and the consumer (R.Buyya, 1999). In a simpler way, it can be considered as a virtualization of a data center, where servers are virtualized seeking the best use of their resources that are made available

through virtual machines. A cloud can be deployed publicly, with a service provider providing the resources and services that the organization needs, or privately, and is managed internally by the organization. There may also be a hybrid structure where an organization maintains an internal infrastructure and provides some services publicly. A public cloud is characterized by being available through a third party service provider via the Internet. It is a cost-effective way to deploy an IT solution, especially for small and medium-sized enterprises, and government entities that need to provide a variety of services to the population. The leading companies that provide cloud computing services are Amazon AWS, Microsoft Azure and Google Cloud. In addition to these, there are several companies that provide only certain services, such as storage.

### 3 OBJECTIVES

The objective of the present work is to evaluate the potential of the Google Earth Engine Explorer tool aiming the generation of thematic maps for the delimitation of urban areas using Cloud Computing, Big Data and Remote Sensing.

- To evaluate the mapping of urban areas from the Google Earth Engine platform using Explorer mode, with the CART classifier, to the limits of the Rio de Janeiro city and scale regional too.
- Generate a mapping of urban delimitation and expansion considering different occupancy intensities.

## 4 CONCEPTUAL REVIEW

### 4.1 Data and Spatial Information

The given terms and information academic are used several times with the same meaning, but in reality they have different meanings. Given is the set of different observations that are collected and stored. Information is a data that is useful for answering questions or for solving a problem, and has a interpretation.

Two major trends have had a profound impact on spatial data management in recent years. The first of these is the exponential growth of the data volume.

the second is the change of criteria for a database to be considered to be large, and this is the result of the wide integration of spatial information in productions continues from many users.

Another important aspect is the current spatial distribution of the users of geographic information accessing contents in different places and with differentiated platforms.

### 4.2 Geographic Data Base and BIG DATA

A geodatabase is a database capable of storing, querying and manipulating geographic information and spatial data of any kind.

With it you can manipulate simple geometric shapes: dots, lines and polygons, or to use features such as topology, 3D technology, images, to better represent real-world phenomena.

The geodatabase can be stored on top database management systems and allows to organize and manage a huge volume of high-performance data in a multi-user environment.

The challenges in a geodatabase, among others, are: analysis, capture, treatment, search, sharing, storage, transfer, viewing and data privacy information.

Big Data is a term that has long been used to name very large or complex data sets that traditional (desktop) data processing applications still can not handle.

In other words, the ever-increasing production of large-volume data requires new alternatives for data management, online data sources, and the need for processing beyond traditional methods.

Cloud computing distributes resources in the form of services. We can then divide it into three types: Software as a Service (SaaS), Platform as a Service (PaaS) and Infrastructure as a Service (IaaS).

Software as a Service (SaaS) The SaaS model provides software systems for specific purposes, which are accessed via the Internet by a web browser for example. In SaaS, you do not manage or control the infrastructure of the lower layers, except for specific system configurations. With this, developers focus on innovation rather than infrastructure, leading to the rapid development of software systems (Souza et al., 2010). user can use vendor applications that run on a cloud infrastructure. But the consumer does not manage or control the basic infrastructure, including network clouds, operating systems, storage, servers. SaaS

runs entirely on the cloud bringing cost savings, thus dispensing with the acquisition of software licenses.

Platform-as-a-Service (PaaS) PaaS providers provide development environments so that developers do not have to worry about the infrastructure that will be used, not even with the installations of the environments used by their applications, expensive and complex jobs in the vast majority of cases. In PaaS, users do not have cloud infrastructure management, but have control over the deployed applications and the ability to configure the hosted application environment. consumer can deploy in the cloud the infrastructure created or the acquired applications created using programming languages and the tools supported by the provider. It does not manage or control the basic infrastructure, but controls used applications and hosting applications and environment settings. Provides an infrastructure to deploy and test applications in the cloud. It also provides an operating system, programming languages and development environments for applications, aiding the implementation of software, as it contains development tools and collaboration between developers.

Infrastructure as a Service (IaaS) the term IaaS, refers to a computational infrastructure that uses virtualization techniques to deliver computational resources. An infrastructure in the IaaS model aims to make it easy and affordable to manage and deliver computing resources, that is, it is responsible for providing resources such as servers, network, storage and even operating systems and applications required to build an environment on demand. In addition to providing in most cases, online services for infrastructure administration, such as a web interface. Because it is the lower layer, it is also responsible for providing the infrastructure used by the middle and upper layers. Amazon EC2 (Cloud Virtual Servers) and Amazon S3 (Scalable Cloud Storage) services are examples of IaaS. Cloud providers usually charge for the IaaS service for the total of resources allocated or consumed (Amazon AWS, 2016). brings the services offered at the infrastructure layer, in these services we can include various computing resources such as servers, routers, storage systems, among others. Responsible for providing all the necessary infrastructure for the two previously mentioned structures SaaS and PaaS. IaaS is based on computing resource virtualization techniques. Looking from the economy side, haven't to buy new servers or even network equipment for service expansion, because everything that is need is included in the cloud.

### 4.3 Urban Areas

The process of urbanization in Brazil began in the twentieth century, and the industrialization led the population to go from the countryside to the city (rural exodus).

Urbanization is an increase of the urban population compared to rural. The economy in Brazil went from agrarian-exporter to urban-industrial, and the understanding of the urban is still based on the idea of capitalist and industrial society.

Currently, more than 80% of the Brazilian population lives in urban areas.

Disorganized urbanization causes a series of social and environmental problems.

The solution to environmental and social problems is what motivates today the study of the growth and densification of urbanization.

### 4.4 Google Earth Engine and the Explorer Platform

Google Earth Engine is a cloud computing platform for developing and hosting web applications on Google's infrastructure that enables, among other applications, automatic mapping and classification of land cover.

It was initially released as a beta release in April 2008 but only in 2017 there was a massive publicity and incentive to use by the company (in Brazil). Pesquisando Google Earth Engine no Google acadêmico podemos achar 20 resultados de artigos ou publicações no período de 2008 a 2016 e 26 no período de 2016 a 2017.

Google Engine is a technology in the Platform as a Service (PaaS) model, which virtualizes applications on multiple servers, providing hardware, connectivity, operating system, and software services. It can be used for free to a certain level of resource consumption with numerous tools ready and adapted for users without programming knowledge. Additional fees may be charged for the consumption of resources when the user has technical knowledge to implement more complex analyzes.

The rapidity in the classification and generation of products is one of the most important positive points of the analysis. The tool is very interactive and easy to handle, even by users with little experience.

It uses cloud-processing technology, which enables the use and manipulation of large volumes of georeferenced data. Its system is designed to enable scientific analysis and visualization of

geospatial data sets. The data catalog has an immense volume of data and a wide range of popular datasets, such as the world's largest collection of Landsat scenes, 25 years of high-resolution images, and other Landsat images since 1972, has variety of data types, bands, projection, bit depth, spatial resolution, temporal. It has data from the Sentinel, Images Moderate Resolution Imaging Spectroradiometer (MODIS), night-time imagery-Defense Meteorological Satellite Program's Operational Linescan System (DMSP-OLS), digital elevation models, slope data, surface temperature, climate, atmospheric data beyond of global daily satellite feeds.

The user can add and store their own data and collections. Upload your own image with Maps Engine, in the original projection, with all the bands and metadata. In addition to being able to save your data, collected points, classifications and these can be used, manipulated and opened in traditional programs.

The platform can be used in two ways, in the "Explorer" mode a programming interface that was used in this work, also contains the "Code Editor" mode used by users with programming knowledge.

The platform has already been successfully used for various purposes. Between them:

The European Commission's Joint Research Center (JRC) has used the Earth Engine to develop high-resolution maps of global surface water occurrence, change, seasonality, recurrence and transitions.

Collect Earth, developed by the Food and Agriculture Organization of the United Nations (FAO), is a free, open-source, easy-to-use tool using Google Earth and Google Earth Engine to view and analyze land lots to deforestation and other forms of land use change.

Global Forest Watch, an initiative of the World Resources Institute, is a dynamic online forest monitoring system designed to enable better management and conservation. Global Forest Watch uses Earth Engine to measure and visualize changes in the world's forests.

A team led by Matt Hansen of the University of Maryland used the Earth Engine to research more than a decade of global warming extension, loss, and gains. This area is 128.8 million square kilometers, equivalent to 143 billion pixels of Landsat data in a spatial resolution of thirty meters.

## 5 METHODOLOGY

The proposed methodology starts with the choice of analysis scales and inputs. The CART classifier for the mesoscale Rio de Janeiro was initially evaluated. Simplified legends (urban and non-urban) and greater detailing (different types of coverage of urban areas and levels of urban intensity) were tested. The main input was the Landsat TOA (Top of atmosphere) mosaic. The potential, time of classification, and results were evaluated.

Using the Google Earth Engine platform in "Explorer" mode for computer laymen and selecting images from different years in the platform database, the classifier was chosen. In this case the CART-Classification and Regression Trees (The decision tree method is a supervised learning approach, that is, it comprises the abstraction of a knowledge model from the data presented in the form of ordered pairs (desired input and output) [Goldschmidt e Passos 2005]. In this method, the production of the results presents simplicity and readability for its interpretation, fact that, according to Oliveira (2005), has become one of the main advantages of its use. With regard to the CART algorithm, one of its main characteristics is the research capacity and relations between the data, involving the construction and simplification phases of the decision tree, choosing the best variable for dividing the data into two nodes, where the procedure is applied recursively to the data in each of the child-nodes and so on [Hand et al 2001].) (a classification algorithm that has one of its main characteristics the research capacity and relations between the data, involving the phases of construction and simplification of the decision tree, choosing the best variable for dividing the data into two nodes, where the division procedure is applied recursively to the data in each one of the nodes (classification is a process that finds common properties between a set of records belonging to a database and classifies them into different classes according to a model) through the indication of samples pixel-by-pixel of the different types of subtitles that were used.

Five tests were performed. In the first test we tried to define the potential of delimitation of urban areas considering only two classes (urban and water) and others. In the second test the number of classes was increased. were selected for vegetation, soil and sand. In this case, the results were more promising.

In the third test, the classification for the delimitation of urban areas on a regional scale for the southeastern region of Brazil was enhanced. In this case the Night time light file was used as support.



Version 4 of the DMSP-OLS Nighttime Lights Time Series consists of cloud-free composites made using all the available archived DMSP-OLS smooth resolution data for calendar years. In cases where two satellites were collecting data - two composites were produced. The products are 30 arc second grids, spanning -180 to 180 degrees longitude and -65 to 75 degrees latitude.

In the final tests (4 and 5) classifications were made to delimit different intensity levels of urban areas. This evaluation considered the scale of detail (minutemen of Rio de Janeiro10 and carried out for the years 2000 and 2010, aiming at the use in the monitoring.

After the generated products was applied Kappa (Cohen, 1960) Gong & Howarth (1990) used the Kappa (K) index as an important precision measure to be associated with the error matrix, since it represents it entirely, that is, considering all elements of the matrix and not only those that are on the main diagonal, as it happens with the Global Accuracy index. The index using an image of Google Earth itself and 200 random points to make the conference of the classified areas. The Kappa index is a pointer to show reliability of the measurement procedures used which is a key issue in any research study. It is a measure of agreement that measures the degree of conformity beyond what would be expected by chance alone. And with all the finished products, the results of the images, classifications (fig.1 and 2), tables 1 and 2, and evaluating the Kappa index that was calculated as shown by the equation below can be analyzed.

$$k = \frac{\sum x_{ij} - \sum x_{i.} \sum x_{.j}}{1 - \sum x_{i.} X_{.j}} \quad (1)$$

Equation 1: Kappa equation.

Table 1: Kappa value and index of points.

Kappa value	Class
< 0.20	Poor
0.21 - 0.40	Low
0.41 - 0.60	Moderate
0.61 - 0.80	Good
0.81 - 1.00	Very good

Font: Adapted from Galparsoro and Fernandez (2001).

## 6 RESULTS AND CONCLUSIONS

### 6.1 Results

The table 2 shows the results to the tests 1 -4.

Table 2: Results to the tests 1 and 2.

Tests	Image/mosaic/ area and scale	Class and number samples	Results
(1)	LANDSAT 8 Mosaic year (2000) - Rio de Janeiro city 1:100000	Urban (83), Water (71), Other (31)	Few class the result overestimate the urban areas
(2)	LANDSAT 8 Mosaic year (2000) - Rio de Janeiro city 1:100000	Urban (83), Water (71), Vegetation (66) Soil (64) Beach sand(22) Other (10)	Delimits urban areas, water bodies, vegetation with good results.
(3)	LANDSAT 8 Mosaic year (2000) + Night time Lights – regional scale 1:1000000	Urban(83), Water(71), Vegetation (66), Soil (64), Beach sand(22) Other (10)	Better delimitation of the urban areas.
(4)	LANDSAT 8 Mosaic years (2000, 2010) - Rio de Janeiro city 1:100000	Intense Urban (20), Average urban (22), Rarefied urban (24), Vegetation (08), Water (18)	The result delimits well diffetents urban levels.

It is possible to observe through the figures 1 and 2 of the generated classifications the following results.

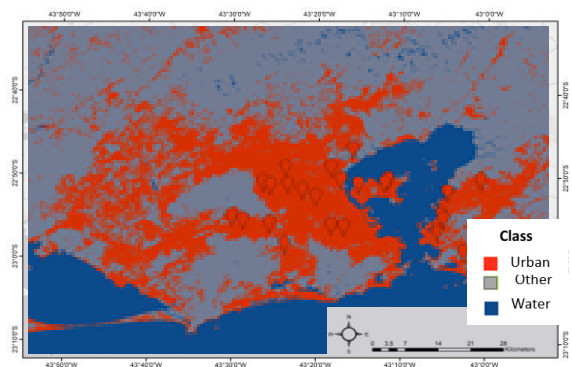


Figure 1: Simplified classification (test 1).

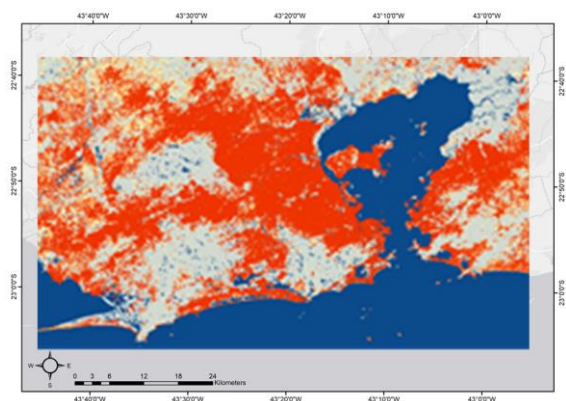


Figure 2: Detailed classification (test 2).

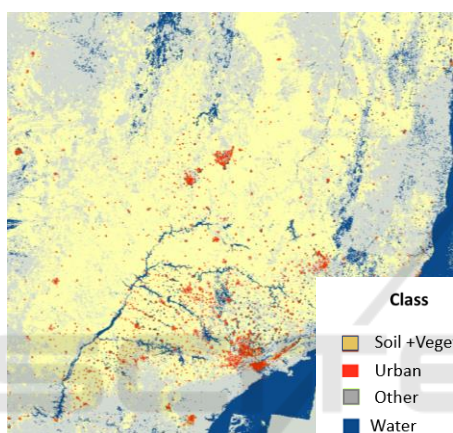


Figure 3: Nigth time ligh classification - Image index Kappa de 0,85 (test 3).

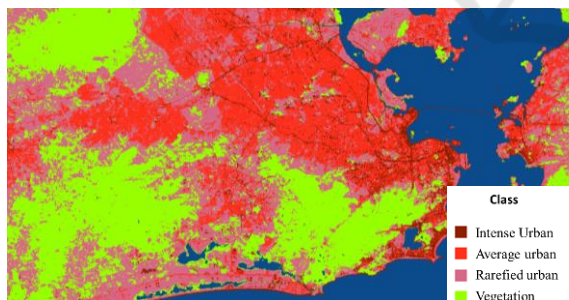


Figure 4: Image and classification of the Rio de Janeiro city in 2000 (test 4/5).

## 6.2 Conclusions

About the Google Earth Engine Explorer Platform was observed high speed in the processing of the samples and in the generation of the result (mapping of the cover); allows to open different tabs for the generation of differentiated classifications; the number of classes and samples very little affect the

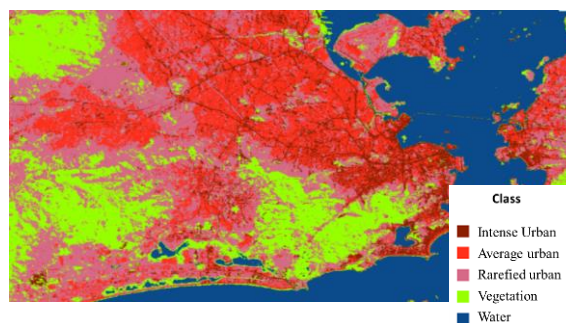


Figure 5: Image and classification of the Rio de Janeiro city in 2010 (test 4/5).

processing speed which is very high; changing the resolution to the classification changes the result and must consider the scale of the mapping; and the classifier proved efficient for the delimitation of urban areas.

On the mapping one can conclude that its use has great potential for Geography with excellent result of the Kappa index; it has the possibility of applications aiming the monitoring of the urban expansion; and the possibility of geographical analysis at different scales.

It is necessary to evaluate the other classifiers as well as to compare tests in other scales and geographic areas.

Validation must consider different methods and validate other images made available on the Google Earth Engine Explorer platform.

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