

DETECTION OF POTATO PLANTATION AREAS WITH MEDIUM RESOLUTION IMAGES: CASE OF BUENO BRANDÃO, MG.

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ABSTRACT:

In the city of Bueno Brandão, South of Minas Gerais State, Brazil, the Watershed of Rio das Antas is located prior to the public water supply and is susceptible to hydro-degradation due to the intensive agricultural activities developed in the area. The potato plantation is the most significant cropping in the city. Because of the possibility of interfering in the preservation areas, mainly the ones surrounding water courses and springs, it is very important to do the assessment of the plantation sites, in order to avoid the risk of water contamination. The procedures adopted by the agro activity farmers generally present the following features: intensive use of agro-chemicals, cropping in places with slopes which are higher than 20%, close to or in permanent preservation areas. The scope of this study was to develop the proper methodology for the assessment of the plantation areas, regarding the short time of procedure, as the period between the plantation and the harvest occurs in six months the furthest. These areas vary year in year out, as the plantation sites often change due to the land degradation. Because of that, geotechnologies are recommended to detect the plantation areas by the use of satellite images and accurate data processing. Considering the availability of LANDSAT medium resolution images, methods for their appropriate classification were approached to provide effective target detection.

1. INTRODUCTION

According to IBGE (2007), the agricultural activities with the most number of producers in Bueno Brandão, MG, are the potato and coffee plantations. However, the intensive use of agricultural defensive chemicals, the elevate frequency of planting, the presence of land disproved of vegetation protection and the plantation in preservation areas, made the potato cropping highly responsible for potentializing the risk of water contamination in the Sub-basin of Antas River. However, prior determining the sites with a major risk of contamination, it is necessary to develop an effective methodology for detecting plantation areas, which is the main scope of this study.

On one hand, the potato plantation must be held in dry sunny places, with a reasonable availability of water for irrigation. On the other hand, the area must present easy drainage features, as the plant does not tolerate water excess. Because of that, Bueno Brandão has been the ideal place for this culture along history. It represents a mountainous area with an abundant amount of water.

For the correction of soil, fertilizers are generally applied before and after plantation. In Bueno Brandão, as most of the areas used for plantation have a declivity higher than 20%, there should be adopted soil conservation methods, minimizing the superficial flow of water, which carries all materials disposed on the land to the rivers, as there is mostly no natural vegetation protecting them. In spite of that, traditional and non-conservationist methods are commonly applied in environmentally fragile areas of plantation.

In that area, the predominating agricultural model may cause degradation to the water resources, mainly because of: trespassing preservation areas around springs and water courses, destroying the vegetation covering plantation sites, and using excessive and inadequate chemicals. According to Boulomytis (2008), some of the most used chemicals by the producers are: Astro, Bravonil, Curzate BR, Dithane PM and Karate Zeon 50CS, which are classified from dangerous to potentially dangerous in the toxicological and environmental hazard patterns (ANVISA, 2008). The most present metals in the formula of these defensive chemicals are cadmium, cobalt, chromo, manganese and zinc.

According to CONAMA Resolution n.303 (2002), there are many kinds of places considered as preservation areas. However, the approached areas are the most relevant ones for the study, located: a) in the river margin, measured from the highest level, in horizontal projection, with the minimum width of 30m, for the water course of less than 10m wide; b) around any kind of water spring, with the minimum radius of 50m; c) in a declining area superior to 45°, at the line of highest declivity. In all cases, according to CONAMA Resolution n.369 (2006), the intervention or suppression of vegetation in preservation areas must not compromise their environmental functions, specially: a) the stability of the margins of the water courses; b) the drainage and the intermittent water courses; c) the maintenance of the biota; d) the regeneration and maintenance of native vegetation; e) the water quality.

For the assessment of plantation areas, it is important to consider that, in the dry season period, August to September, the preparation of the land occurs, making it more appropriate to visualize the bare soil areas by the use of satellite images, and accurate techniques to provide the target detection.

2. MATERIALS AND METHODS

2.1 Study area

The city of Bueno Brandão is located in the south of Minas Gerais State, in the micro region of Alta Mantiqueira. The highest altitude is 1,719m and the lowest is 840m, with S 22°26'27" latitude and W 46°21'03" longitude, of W 45° central meridian and fuse 23. The region has rivers with a low discharge volume, among which are the rivers called Antas, Cascavel and Cachoeirinha. The local topography is mountainous in 70% of the area, with the presence of many waterfalls. The municipal area is 355.23 sq-km, comprehending the watershed of Rio das Antas. It is located prior to public distribution and has the area of 50.22 km², with latitudes between S 22°31'38" and S 22°24'15" and longitudes between W 46°22'13" and W 46°13'56" geographical coordinates (Figure 1).

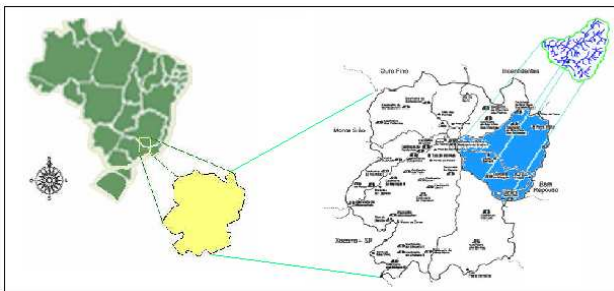


Figure 1. Location of the study area, highlighting the municipality of Bueno Brandão (on the left) and the watershed of Rio das Antas (on the right) in the State of Minas Gerais, Brazil (Boulomytis, 2008).

The mountainous condition of the area might be represented by different classes of slope, according to the approach of Vieira (1988). They describe the natural limitation to use the land, regarding the agro plantation. The definitions of the six classes are detailed as it follows: Class A) soft slope or practically plan, inferior to 3%; Class B) low slope and slightly hilly, between 3% and 8%; Class C) moderate slope, between 8% and 20%; Class D) severe slope, between 20% and 25%; Class E) very severe slope, between 45% and 75%; Class F) cliffs, with slopes superior to 75%.

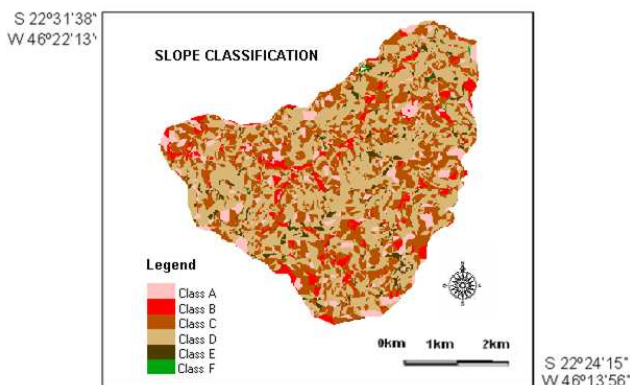


Figure 2. Thematic map of slope, classifying the natural limitation of the Watershed of Rio das Antas for the agro plantation use (Boulomytis, 2008).

The condition which characterizes the area (Figure 2) justifies the importance of the study, based on the fact that, even though the area has a natural limitation to the agro plantation it is still intensively used for this purpose.

2.2 Materials

For the study, a LANDSAT TM5 medium resolution image with the following features was used: orbit/point 219/76, bands 3,4 and 5, from August 16th, 2007, available at the DGI-INPE Home Page.

http://www.dgi.inpe.br/siteDgi/index_pt.php. For the registration of the image, the Chart of Brazil – Munhoz and Ouro Fino, in the scale of 1:50.000, was used (IBGE, 1972).

The classification of the area was done by the use of: 1) the SPRING v 5.1.8 software (Câmara et. al, 1996), for the pixel-based classification; 2) and, the Definiens Developer® platform, for the OBIA (Object-Based Image Analysis).

The total of 100 land use and cover samples were collected in the field with a GPS (Global Position System) receptor (30 samples of potato plantation sites) and with the ©2009 Google Earth – Image © 2010 Geoeye (20 water courses, 30 other-plantation areas and 20 forest areas).

2.3 Methodology

The methodology of the study had the purpose of detecting the potato plantation areas by the use of the most effective procedure. Thus, two techniques were approached and compared: 1) the pixel-based classification; 2) the OBIA.

The reason for achieving the image in this period of the year (August to September) is that there is bare soil and lack of vegetation in the potato plantation sites, making it visually easier to identify this cluster.

Regarding the pixel-based classification technique, for the segmentation, three processes were done: 10x10, 10x15 e 10x20 (10 tons of gray per region of 10, 15 or 20 pixels). After analyzing the most appropriate result, the segmentation of 10x10 pixels was selected, in order to gather the cluster with fewer cells and more similar tons of gray. This procedure promoted a better pixel-based classification afterwards. The classifier used was Bhattacharyya, with a limier of acceptance of 99%.

Samples were used for training the classification algorithm, based on the distribution of probability in each class. They contemplated 10 samples of each kind. It was possible to identify an area of representative and homogeneous sampling, including the variability of gray levels for each class (Câmara et al., 1996).

The OBIA was also based on the segmentation and classification steps. At the first step, objects were created in different scales, according to the criteria of shape, color and homogeneity, all connected. At the second step, the objects became related by the definition of a class hierarchy and semantic information. In the image, an object represents an entity which might be individualized by the class attributes and properties of the original data. These attributes do not only correspond to the spectral feature of the objects, but also to the topological relationships, texture, shape, and size, among others. The segmentation is a fundamental step in this process, responsible for generating spectrally homogeneous segments

which represent the inherent dimensions concerning the objects contained in the images (Blaschke, 2010).

In the OBIA process, a data bank was created in the Definiens Developer® Plattform, containing the Landsat TM5 image and the surrounding limit of the watershed. Based on previous researches (Alves *et al*, 2010 and Alves *et al.*, 2009), and considering the characteristics of the used data and the purpose of the classification, parameters of scale, shape and color were defined for the segmentation. Initially, two levels of segmentation were created:

- 1st level - algorithm *multiresolution segmentation*, scale parameter 10; shape 0.1 and compactness 0.5; weight 2 for band 4, and weight 1 for the remaining bands. This segmentation provided the level called MRS10 with 8.365 objects;
- 2nd level – algorithm *spectral difference segmentation*, spectral distance of 6. This segmentation provided the level called SDS6 with 3.244 objects.

For the validation issue, the samples which had not been used in the supervised classification were later processed (20 potato plantation areas, 10 water courses, 20 other-plantation areas and 10 forest areas). By the use of the Kappa index (Congalton and Green, 2009), a confusion matrix of the classification was done and the Z test processed for both of the techniques.

3. RESULTS AND DISCUSSION

The classes were attributed as the following: potato, other plantations, forest and water courses (Figure 3).

3.1 Pixel-based classification

Even though the region has abundant hydric resources, the “water courses” class corresponded only to a little more than 0.6% of the area. It occurred because, from orbital images, there are water courses that may not be detected due to the vegetation which covers them and the mountains surrounding the area. It becomes difficult to obtain the drainage line only by applying traditional classification methods. As it was not the purpose of this research whatsoever, no specific methodology to satisfy this demand was processed.

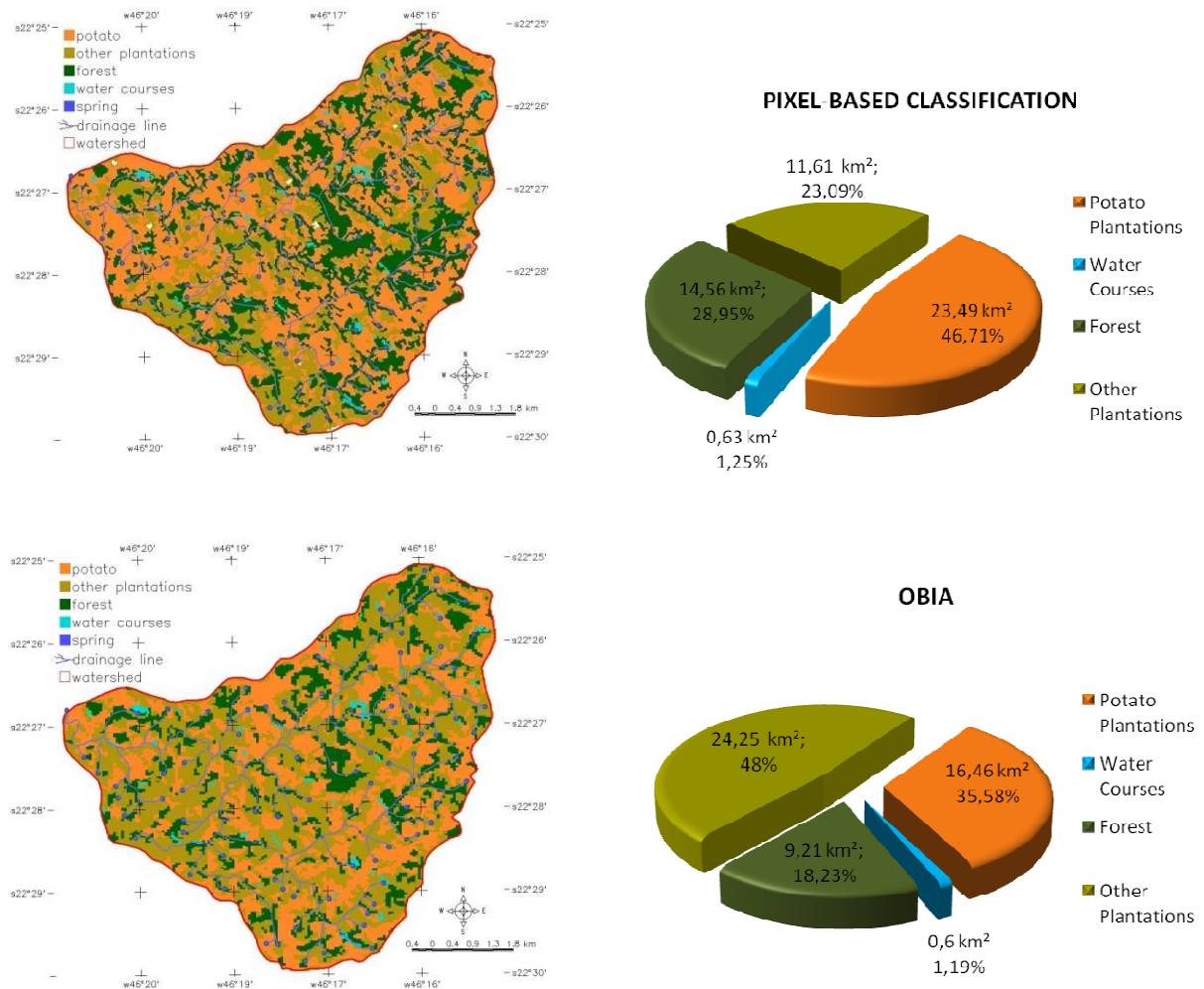


Figure 3. Classification of land use and cover of the Watershed of Rio das Antas, Bueno Brandão, MG, Brazil, using both of the approaches: Pixel-based Classification and OBIA.

It is possible to verify the presence of vegetation over some drainage lines displayed at Figure 4. Nevertheless, it was important to do the detection of watercourses in order to obtain the area with a suppression of riparian forests along rivers and springs.

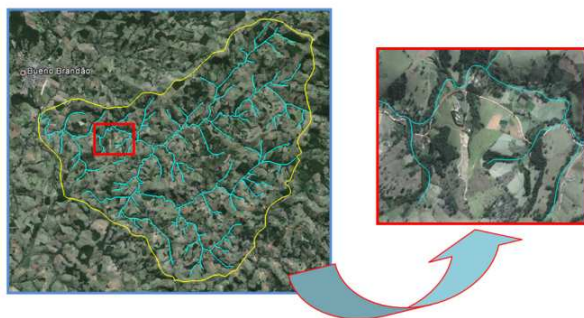


Figure 4. Samples of drainage lines covered by vegetation in the Watershed of Rio das Antas, Bueno Brandão, MG, obtained by the use of ©2012 Google Earth – Image © 2012 Geoeye.

The vegetation areas corresponding to other kinds of agro plantations and pastures were classified as “other plantations”. These other kinds of agro plantations, such as coffee and corn, represent a smaller percentage than the potato plantation. Most of the “other plantations” class is represented by pastures, as the predominant system of managing the cattle in the region is extensive and depend on large areas for feeding the animals. As it was not the scope of the article, there was no aim to classify the pasture individually. However, this result represented some incoherence at the classification, because even though Bueno Brandão is commonly one of the most significant producers of potato in that region, it should not comprehend the total amount obtained in this area.

According to IBGE (2007), in 2006 the production of potato in Bueno Brandão was in 10,20 km², which would represent less than half of the classified area. One of the main reasons for the occurrence of this pattern in the classification is probably because there was a clear confusion between the other bare soil areas, which did not correspond to potato plantations. As there was not a different attribute to distinguish between the bare soil areas regarding potato plantation and other plantations, all of them were classified the same way.

3.2 OBIA

The first level of segmentation, MRS10, generated a high level of separability and detailing, allowing the separation of small objects. The higher weight given to band 4 was used to highlight the vegetation response in the near infrared, distinguishing the vegetation areas (forest and other plantations), which presented bare soil.

The algorithm *spectral difference segmentation* was used to assemble contiguous objects with similar spectral characteristics. In this case, the used parameter considered up to 6 levels of gray. Thus, the small objects generated in the MRS10 level were preserved, preserving also those which represented small potato plantations. This algorithm allowed the reduction of object numbers, reducing the processing time of classification and minimizing the internal details of non-desirable thematic classes. In the SDS6 level, the final classification was finally done.

As it may be observed on Table 1, the attributes used were spectral and relational. It was not possible to use shape attributes due to the lack of regularity and reduced dimensions of potato plantation sites, besides the presence of slopes in the same region where they are located.

Class	Attributes	Attributes types
Water bodies	NDWI	spectral
Forest	NDVI	spectral
Potato plantations	NDVI	spectral
	max. difference	spectral
Other plantations	Not potato plantation	relational
	Not Forest	relational

Table 1. Attribute and attribute types used for the OBIA classification technique.

The results obtained for the potato plantation were more effective, considering the estimation of IBGE (2007), although there was also a significant variation from the survey. However, analyzing the consistence of proportion of the other classes, the vegetation areas were preserved and properly distinguished from the potato plantation areas.

3.3 Validation process

The confusion matrix was generated after crossing the samples with the reference map and both of the applied techniques. The results may be seen on Table 2.

	Global Accuracy	Kappa index	Z test ($\alpha=5\%$)	Comparative Z test ($\alpha=5\%$)
Pixel-based Classification	0,62	0,47	5,19	- 1,97
OBIA	0,78	0,70	9,47	

Table 2. Global Accuracy, *kappa* index and Z tests obtained for the classification using both techniques.

The classification with the pixel-based approach obtained a theme map whose global accuracy and *kappa* (K) indexes have been respectively 62% and 0,47. The value found for *kappa* coefficient was acceptable but close to poor ($K < 0,40$), based on Congalton and Green (2009). It showed the lack of conformity of the classification compared to the reference map. The Z test applied for the image classifications indicated that the null hypothesis, considering the classifications to be similar to a random classification, was negated, since the kappa values were significantly higher than zero, for the significance level of 5%. The classification with the OBIA approach obtained a theme map whose global accuracy and *kappa* indexes have been respectively 78% and 0,70. The value found for *kappa* coefficient was acceptable ($0,4 < K < 0,8$), based on Congalton and Green (2009). The classification was more proper when compared with the reference map. It was confirmed by the Z test in which there is an agreement between classifications and the reference image, for the adopted level of significance ($\alpha=5\%$). The result of it for the pixel-based classification was 9,47.

The Z test was also applied in a comparative way to verify the results of the two methods. The result obtained, at a significance level of 5%, confirms that the value of Kappa by the traditional method (pixel-based classification) is significantly lower than that obtained by the second method, OBIA.

4. CONCLUSION

The validation process showed that the approach using OBIA was more accurate, as it also considered relational attributes, beyond others.

It is important to notice that the methodology using the pixel-based classification might be applied only when the image is in the proper period, while the farmers are preparing the land and there are bare soil areas. In case the potato crops area growing and flourishing, the OBIA might be a more appropriate technique, because besides the spectral and texture attributes, relational attributes may also be used, considering the same classes in different dates, such as, the object attributes, the relationship between classes, the global relationship and the logic operators (that may occur at the same class hierarchy level or at higher or lower ones).

Finally, this methodology proved to be very useful for the assessment of potato plantation areas prior to the actual plantation, while the producers are still preparing the land. In case it is in an environmentally fragile area, there is still a possibility of avoiding the continuation of plantation by the local authorities, minimizing the risk of water contamination.

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REFERENCES

Agência Nacional de Vigilância Sanitária - ANVISA. Sistema de Informações sobre Agrotóxicos - SIA. <http://www4.anvisa.gov.br/agrosia/asp/default.asp> (accessed 08 Jan. 2008)

Alves, C.D., Pereira, M. N.; Florenzano, T. G. ; Souza, I. M. E., 2009. Análise orientada a objeto no mapeamento de áreas urbanas com imagens Landsat. *Boletim de Ciências Geodésicas*, 15, pp. 120-141.

Alves, C.D., Florenzano, T. G., Pereira, M. N., 2010. Mapeamento de áreas urbanizadas com imagens Landsat e classificação baseada em objeto. *Revista Brasileira de Cartografia*, 62 (2), pp. 189-198.

Blaschke, T., 2010. Object based image analysis for remote sensing. *ISPRS Journal of Photogrammetry and Remote Sensing*, 65(1), pp. 2-16.

Boulomytis, V. T. G., 2008. *The use of geo-technologies to evaluate the potentiality of hydro-degradation in superficial waters by agro-chemicals: case of Antas River watershed, Bueno Brandão, MG*. Thesis (Master's Degree in Civil Engineering). Universidade Estadual de Campinas, Campinas.

Câmara, G.; Souza, R.C.M.; Freitas, U.M.; Garrido, J.; Li, F. M., 1996. SPRING: Integrating Remote Sensing and GIS by Object-Oriented Data Modelling. *Computer and Graphics*, 20 (3), pp.395-403.

Congalton, R.G, Green, K., 2009. *Assessing the accuracy of remotely sensed data: principles and practices*. Taylor and Francis, New York, 183 p.

Conselho Nacional de Meio Ambiente – CONAMA. 2002. Resolution n. 303. http://www.ambiente.sp.gov.br/uploads/arquivos/legislacoesambientais/2002_Res_CONAMA_303.pdf (accessed 10 Feb. 2012)

Conselho Nacional de Meio Ambiente - CONAMA 2006. Resolution n. 369. http://www.ambiente.sp.gov.br/uploads/arquivos/legislacoesambientais/2002_Res_CONAMA_369.pdf (accessed 10 Feb. 2012)

Definiens Developer® Plattform. eCognition: User Guide 5, 2006. http://www.definiens.com/services/faq/Definiens_Professional5_InstallationGuide.pdf (accessed 02.mar.2010)

Divisão de Geração de Imagens – DGI/INPE. LANDSAT TM5, orbit/point 219/76, bands 3,4 and 5, from August 16th, 2007 http://www.dgi.inpe.br/siteDgi/index_pt.php (accessed 18 Mar. 2008)

Google Earth, 2009. Image © 2010 Geoeye. Project coordenates: S 22°31'38'' - S 22°24'17'' and W 46°22'13'' - W 46°13'56'' (accessed 02 Oct. 2009).

Google Earth, 2012. Image © 2012 Geoeye. Project coordenates: S 22°31'38'' - S 22°24'17'' and W 46°22'13'' - W 46°13'56'' (accessed on 03 Mar. 2012).

Instituto Brasileiro de Geografia e Estatística – IBGE, 1972. Carta do Brasil – Munhoz e Ouro Fino. IBGE, São Paulo, IBGE.

Instituto Brasileiro de Geografia e Estatística – IBGE, 2007. Sistema IBGE de recuperação automática (SIDRA). <http://www.sidra.ibge.gov.br/bda/acervo/acervo2.asp?e=v&p=C&A&z=t&o=11> (accessed 10 Feb. 2012).

Vieira, L. S., 1988. *Manual da ciência do solo: com ênfase aos Solos Tropicais*. Editora Agronômica Ceres LTDA., São Paulo, 464p.