

# Spatiotemporal mapping of soybean plantations in Rondônia, Western Brazilian Amazon

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## ABSTRACT

Although soybean production has been increasing in the state of Rondônia in the last decade, soybean planted area has been estimated indirectly using secondary datasets, which has limited understanding of its spatiotemporal distribution patterns. This study aimed to map and analyze spatial patterns of soybean expansion in Rondônia. We developed a classification technique based on Spectral Mixture Analysis (SMA) derived from Landsat imagery and Decision Tree Classification to detect and map soybean plantations in 2000, 2005, 2010, and 2014. The soybean classification map showed 93% global accuracy, 23% omission and 0% of commission errors for soybean crop fields. The greatest increases of soybean cropped area in the state of Rondônia were observed between 2000-2005 and 2005-2010 time-periods (33,239 ha and 59,628 ha, respectively), mostly located in Southern Rondônia. The expansion of soybean areas to Northern Rondônia (25,627 ha) has mostly occurred in the 2010-2014 time period. We estimate that 95.4% of all newly created soybean plantations, detected by 2014, were established on lands deforested nine or more years earlier. We concluded that the incursion of soybean plantations on lands deforested for other land uses (e.g. ranching) is contributing to their displacement (pastures) from older colonization zones toward more remote frontier areas of the Amazon, exacerbating new deforestation there.

**KEYWORDS:** Brazilian Amazon, Tropical deforestation, Land-use and land-cover change, Geoprocessing, Agricultural commodities.

## Mapeamento espaço-temporal de plantios de soja no estado de Rondônia, Amazônia ocidental brasileira

### RESUMO

Os plantios de soja têm aumentado ao Norte e Sul dos municípios de Rondônia na última década, entretanto a área plantada de soja tem sido estimada utilizando dados secundários, o que limita o entendimento da distribuição espaço-temporal da soja. Este estudo buscou analisar e mapear os padrões espaciais de expansão da soja em Rondônia. O mapeamento de plantios de soja nos anos 2000, 2005, 2010 e 2014 foi feito a partir de uma técnica de classificação baseada na análise de mistura espectral de imagens Landsat e em uma árvore de decisão. A acurácia global, erros de omissão e comissão para o mapeamento da soja foram 93%, 23% e 0%, respectivamente. Os resultados mostraram que os maiores incrementos da área de soja ocorreram no estado de Rondônia entre os períodos de 2000-2005 e 2005-2010 (33,239 ha e 59,628 ha, respectivamente). A expansão das áreas de soja para o norte de Rondônia (25,627 ha) ocorreu em sua maioria no período de 2010 a 2014. Observou-se que 95,4% de todos os plantios de soja detectados em 2014 ocorreram em áreas com pelo menos nove anos de desmatamento. Além disso, encontramos evidências de que os plantios de soja estão contribuindo para o deslocamento de usos da terra prévios de antigas zonas de colonização, predominantemente pastos que foram empurrados para outras áreas de fronteiras da Amazônia, e lá exacerbando novos desmatamentos.

**PALAVRAS-CHAVE:** Amazônia Brasileira, Desmatamento, Mudanças no uso e cobertura da terra, Geoprocessamento, *commodities* agrícolas.

## INTRODUCTION

The trade of agricultural commodities plays an important role in the Brazilian economy (Simon and Garagorry 2005). The Brazilian federal government has invested massively to achieve high agricultural production in recent decades, especially in the Amazon region (Chaddad and Jank 2006). Public investments have included funding for improving transport infrastructure to foster the soybean boom (Vera-Diaz *et al.* 2009). Soybean plantations were initially introduced in the Brazilian Amazon in the 1990s after new plant varieties were developed that could tolerate the regional climate (Fearnside 2001).

As a result of these favorable economic conditions, soybean plantations have become an increasing environmental concern in the Brazilian Amazon because of their potential impacts on land use and land cover change patterns (Morton *et al.* 2006; Lapola *et al.* 2010; Arima *et al.* 2011).

In the case of the Western Brazilian Amazon state of Rondônia, the area occupied by soybean plantations has substantially increased in the last decade (Brown *et al.* 2004), with areas cropped with soybean increasing from 118 km<sup>2</sup> to 1000 km<sup>2</sup> between 2000 and 2005 (IBGE 2015). The expansion of soybean plantations is directly related to land availability, relatively lower land prices, high foreign demand for soybean, and increased tax incentives for cultivation of this commodity (Fearnside 2001; Brown *et al.* 2005).

Several studies have used Landsat images to map land use and land cover changes inside areas affected by soybean expansion within the Brazilian Amazon (Brown *et al.* 2005; Morton *et al.* 2006; Mello *et al.* 2012). In addition, Roberts *et al.* (2002) used SMA (Spectral Mixture Analysis) and a decision tree classifier to assess land use changes in three municipalities of Rondônia between 1975 and 1999. Brown *et al.* (2005) assessed land use and land cover change between 1996 and 2001 in the municipality of Vilhena, which is located in of the southern portion of Rondônia. In their research, Brown and his collaborators did not observe significant changes caused by soybean plantation on areas occupied by native forests. Similarly, Brandão *et al.* (2005) suggested that newly deforested land would be not suitable for soybean plantations because of forest leftovers (i.e., wood debris and tree roots) that resist the slash-and-burn process. Moreover, Gollnow and Lakes (2014) found associations between cattle raising, soy expansion and deforestation along the BR -163 in the state of Mato Grosso. These authors also affirmed that cattle ranching had a strong relationship with deforestation until 2004.

In this study we examined the extent of time required for soybean plantations to be established and inferred the potential deforestation leakage effects due to soybean

plantations displacing pastures, many of which appear to be being reestablished by deforesting previously undisturbed forested areas along Rondônia's frontier areas. This study aimed to investigate the soybean spatiotemporal expansion in Rondônia during the period of 2000, 2005, 2010 and 2014.

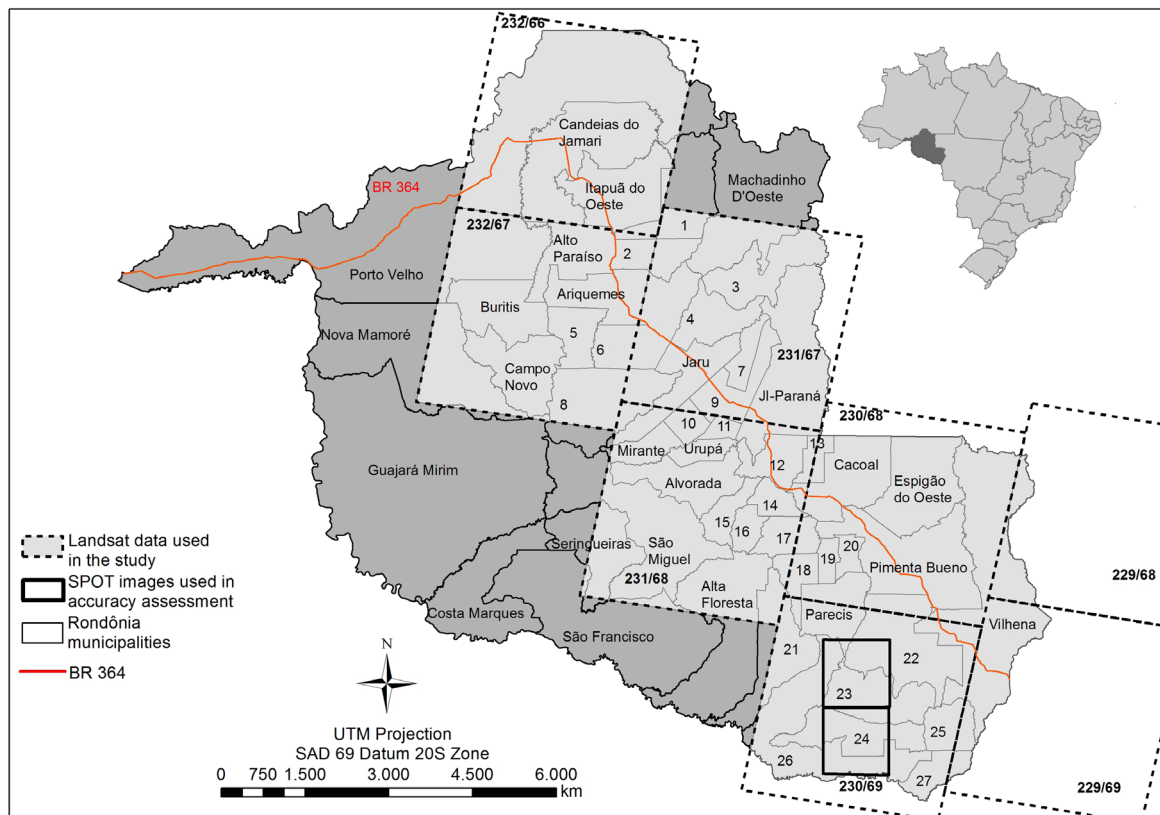
## MATERIALS AND METHODS

Our study area comprised the entire state of Rondônia (~230,000 km<sup>2</sup>) and it included portions of 18 Landsat scenes. Signs of soybean crop lands were searched for throughout the study area. We focused our analysis most strongly in those municipalities that were consistently reported by IBGE (2015) as hotspots for soybean production in Rondônia over the last decade. Evidence of soybean plantations were found within 7 (2000-2005-2010) and 8 (2014) individual Landsat scenes, respectively, which defined the effective area of study used in subsequent analyses (Figure 1).

After our initial assessment, we analyzed the Landsat scenes from four time periods (2000, 2005, 2010, and 2014) that encompassed all municipalities where evidence of soybean plantations existed. The Landsat images used in our analysis were provided by the National Institute for Space Research (INPE). Landsat images were acquired between May and August, the most cloud-free period, which also minimized seasonal effects due to moisture and vegetation growth within the remote sensing data. In addition, we used two SPOT 5 scenes (2.5 meter spatial resolution) and one Landsat TM scene acquired in May 2008 for the accuracy assessment of the soybean classification map (see further details in Table 1).

Fieldwork was carried out in February 2012 to validate the remote sensing techniques and to acquire further land use datasets. This fieldwork included road surveys on the BR-364 federal highway and on the state roads RO-399, RO-370, RO-497, RO-487, and RO-485 for field observations and interviews with government agency employees while searching for supporting data (location and preferred soybean fields, soybean crop cycles, and land prices). The surveys started on February 13<sup>th</sup>, 2012 and lasted ten days. During the fieldwork, we visited and collected *in situ* soybean and deforestation data from sixteen municipalities (Porto Velho, Itapuã do Oeste, Alto Paraíso, Ariquemes, Jaru, Ouro Preto, Ji-Paraná, Presidente Médici, Castanheiras, Rolim de Moura, Pimenta Bueno, Vilhena, Colorado, Corumbiara, Cerejeiras, and Pimenteiras do Oeste).

Radiometric corrections were conducted on the Landsat imagery by converting digital numbers into physical values of radiance and, subsequently, radiance into apparent reflectance at top of atmosphere (Chander and Markham 2003) using ENVI 4.5° software calibration tools. Geometric correction was carried out using orthorectified images acquired from the Global Land Cover Facility (GLCF 2014), University



**Figure 1.** Rondônia municipalities and SPOT and Landsat scenes used to map and verify soybean plantations in 2014 within the state of Rondônia, Brazil. This figure is in color in the electronic version.

1 - Cujubim, 2 - Rio Crespo, 3 - Vale do Anari, 4 - Theobroma, 5 - Monte Negro, 6 - Cacaulândia, 7 - Vale do Paraíso, 8 - Governador Jorge Teixeira, 9 - Ouro Preto do Oeste, 10 - Nova União, 11 - Teixeiraópolis, 12 - Presidente Médici, 13 - Ministro Andreazza, 14 - Castanheiras, 15 - Nova Brasilândia do Oeste, 16 - Novo Horizonte do Oeste, 17 - Rolim de Moura, 18 - Santa Luzia D' oeste, 19 - São Felipe D' oeste, 20 - Primavera de Rondônia, 21 - Alto Alegre do Parecis, 22 - Chupinguaia, 23 - Corumbiara, 24 - Cerejeiras, 25 - Colorado do Oeste, 26 - Pimenteiras, 27 - Cabixi.

**Table 1.** Landsat and SPOT imagery used in this study.

Path/row	2000	2005	2008	2010	2014
MI_1913**			May 25*		
MI_1964**			May 25*		
230/069	July 08	June 04	May 27*	June 18	July 31
229/068	July 17	July 15		June 27	August 09
229/069	July 17	August 16		June 27	August 09
230/068	June 06	July 22		July 20	August 16
231/067	June 29	July 13		June 25	August 23
232/066	July 06	June 02		July 02	June 11
232/067	July 06	June 02		July 02	July 13
231/068					August 23

\* Images used for Accuracy assessment

\*\* Satellite data of SPOT 5

Landsat5 TM data were used for years 2000, 2005, 2008, and 2010 of the study;  
Landsat8 OLI data was used for the year 2014.

of Maryland. The Landsat images were georeferenced using first-order polynomial models and the nearest neighbor resampling technique. Individual geometric correction was deemed acceptable only when RMS was less than 0.5 pixel.

Spectral Mixture Analysis (SMA) was applied to retrieve fraction images (Shade, Non-Photosynthetic Vegetation, Vegetation, and Soil) (eq. 1) used as input in a decision tree classifier from Envi 4.5<sup>®</sup> to detect soybean cropped areas. The spatial mixing proportions of each pixel can be estimated by SMA, which compares mixed spectral signatures of different targets to a group of endmember pixels. This comparison resulted in estimates of different proportions of each endmember within mixed pixels (Adams *et al.* 1995; Lillesand *et al.* 2004). Each pixel is decomposed into a collection of fractions corresponding to the proportion of each endmember present in the pixel (Keshava and Mustard 2002). The endmember pixels were extracted from 2-D scatter plots, using the red and near infrared bands for each scene of Landsat 5 TM and Landsat 8 OLI. Endmembers were collected under similar atmospheric conditions. Field observations of known sites and a visual and spectral examination of spectral shape and image context was conducted accordingly to select the final endmembers for the SMA. Each pixel composition was unmixed into vegetation, bare soil, NPV (non-photosynthetic vegetation) and shadow fractions (Roberts *et al.* 1993). The Linear mixture model is defined by (Rencz 1999):

$$DN_b = \sum_{i=1}^N F_i DN_{ib} + \epsilon_b \quad (\text{eq.1})$$

where:

$DN_b$  = intensity of a given pixel in bandpass 'b'

$DN_{ib}$  = intensity of image endmember 'i' at wavelength 'b'

$R_{ib}$  = reflectance of endmember 'i' in bandpass 'b'

N = number of endmembers

$\epsilon_b$  = error of the fit for bandpass 'b'

Based on RMS (Root Mean Square) matrix of SMA, we discarded models with RMS values greater than 5%. In addition, we inspected histograms of fraction images and only models that presented at least 95% of values within 0% and 100% were kept.

The endmembers used in SMA are a result of mixing with different materials and, therefore, fractional abundance of an endmember may be slightly less than 0 or greater than 1. This conditions indicates that two materials were combined, resulting in a negative or greater than 1 fraction value but those values do not necessarily mean that there are method errors (Rencz 1999).

The mean texture algorithm was used to estimate texture of bands 4 and 7 (Landsat 5 TM) and of bands 5 and 7 (Landsat 8 OLI). For the texture parameters, we applied the filter mean based on occurrence measures in 3 x 3 kernel size. The maximum and minimum values of band 4 and band 5 texture images for soybean areas were defined based on a detailed analysis of the spectral profile of known soybean cropped lands.

Subsequently, we developed a decision tree classification technique using as input the SMA fraction images and texture images to detect soybean cropped areas (Figure 2). We used spectral responses of SMA and texture images as inputs in the different nodes of the decision tree.

First, the vegetation fraction images retrieved from SMA were used to detect forested and non-forested areas, and soil

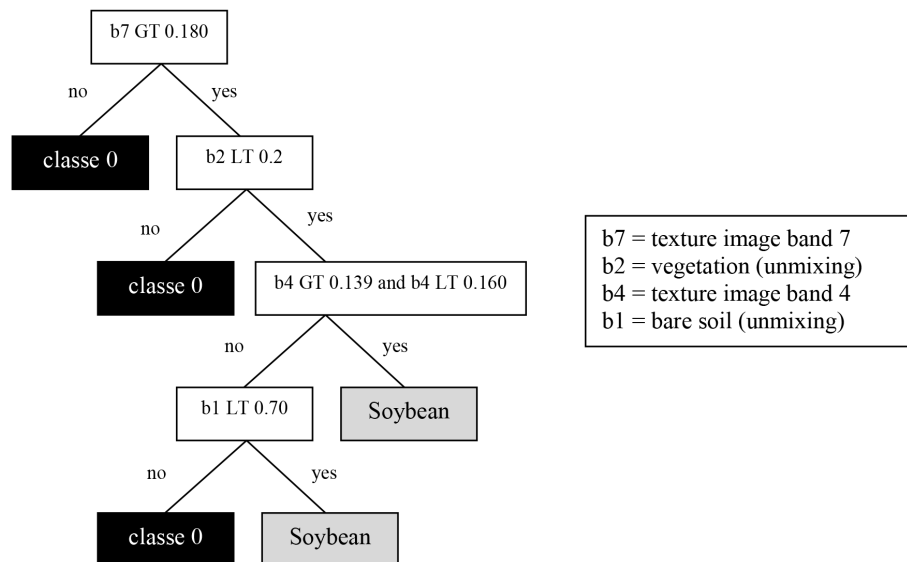


Figure 2. Decision tree used in the study.

fraction images were used to detect soybean areas (fallow areas). Based on our field observations in Rondônia, we assumed that a three month-fallow period (between July 15<sup>th</sup> and September 15<sup>th</sup>) was adopted by the farmers to control plant diseases, following the government policy locally known as “*vazio sanitário*”.

In addition, we used texture images retrieved from bands 4 and 7 for each Landsat scene as an additional input to the decision tree classifier to differentiate pastures from soybean areas. Our field observations indicated that most pasture areas in our study site were formed by a mosaic of grass and isolated trees. The presence of trees created a rougher texture surface on Landsat imagery when compared to soybean cropped areas that are generally free from any trees. Figure 3 shows the different steps used to detect soybean cropped area within our study area.

The accuracy assessment was conducted by comparing binary classifications (1 = soybean cropped areas, 0 = non-soybean cropped areas) from the soybean map derived from Landsat 5 TM imagery with 130 random spatially distributed points which were checked on two higher resolution SPOT 5 scenes. Both Landsat and SPOT images were acquired in May 2008. A confusion matrix developed by Cohen (1960) was used to provide the accuracy quantification of the soybean cropped area classification in this analysis.

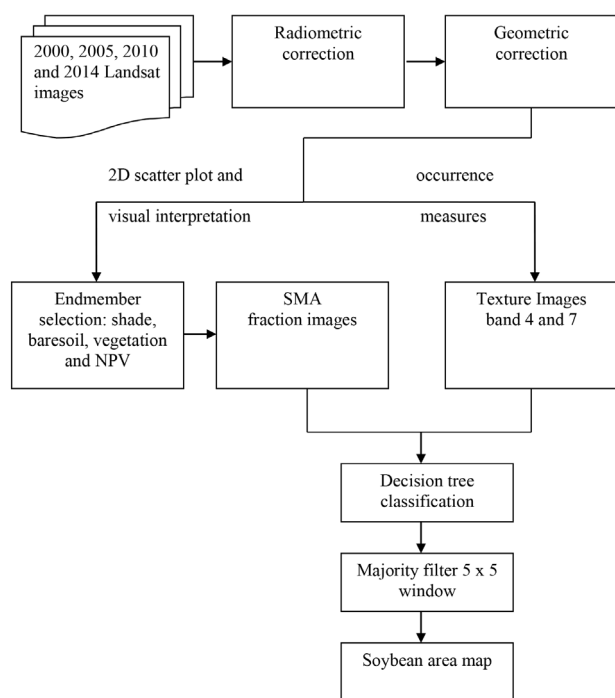


Figure 3. Flow diagram for soybean semi-automated detection technique.

Multi-annual deforestation dataset layers for 1997 through 2014, produced by the Amazonian Deforestation Satellite Monitoring Program (PRODES) at the Brazilian National Institute of Space Research (INPE) were used in this research to identify the year of deforestation for each of the soybean cropped lands. Between 1988 and 2002, the PRODES project used visual interpretation to map deforestation throughout the Amazon. Between 2003 and 2005, PRODES applied SMA and unsupervised classification. Since then, PRODES has been applying the Terra Amazon Classification System and subsequent visual inspection and editing to classify deforestation in the Amazon region. Each annual layer included six classes of land cover types (deforestation, forest, non-forest, water bodies, clouds, residual). Piontekowski *et al.* (2014) have assessed the accuracy of the PRODES dataset for the state of Rondônia and reported an overall accuracy of 93.3% for forest and deforestation classes.

## RESULTS

### Fieldwork

We observed in our fieldwork that farmers till the soils, apply herbicides and remove grasses. Subsequently, in the middle of September, soybean crops are finally planted. The full soybean cycle takes from 100 days (very early-soybean) to 130 days (late or semi-late soybean). Soybean crop harvest begins in January and finishes in March. Just after the end of soybean harvest, a period locally called “*safrinha*” is started. The “*safrinha*” is a practice of planting cover crops of corn or sorghum in the same areas of the soybean harvest. The “*safrinha*” partly overlaps with the period in which soybean planting is not allowed (the “*vazio sanitário*”) which begins on June 15<sup>th</sup> and lasts 90 days. This annual soybean cultivation break has been enforced since 2006 in Rondônia to reduce the spread of diseases in soybeans plantations.

Most of the soybean croplands in Rondônia were observed in municipalities located in the Southern portion of the state (i.e., Vilhena, Corumbiara, Colorado, Cerejeiras, and Pimenteiras do Oeste), but we also observed increasing numbers of plantation areas in municipalities located in the Northern portion (i.e., Ariquemes, Itapuá do Oeste, and Porto Velho) in more recent years. The fieldwork was especially important because it allowed the development of techniques to discriminate soybean and corn areas from pasture within the satellite images.

### Accuracy Assessment

The classifier presents a satisfactory performance with 93% global accuracy, greater than the good accuracy value suggested by Jensen (1986) and Congalton (1991).

Our results showed that the decision tree classification technique correctly detected 77% of soybean areas, without



any commission errors. According to Landis and Koch (1977) our accuracy results indicate a strong agreement with the classified imagery and the high spatial resolution reference image. Furthermore, this technique showed better results for classifying ripe soybeans, immediately before or just after the end of the harvesting season. Table 2 shows further details on the accuracy assessment of the soybean classification maps.

### Soybean fields

The semi-automated soybean classification approach allowed the detection of soybean areas. Most pastures were easily detected because of the presence of palm trees. As a result, NPV fractions from SMA mapping were used to exclude pastures in some cases.

Soybean and pasture areas can be misclassified because of their spectral similarity. The fieldwork was crucial to differentiate soybean from pasture. During the fieldwork we observed that most pasture lands in Rondônia (including those located in the Vilhena municipality) showed presence of palm and other tree species. Based on this feature, we assumed the trees were already present in the pasture areas for 2000 and 2005. As a result, we used the near infrared and shortwave

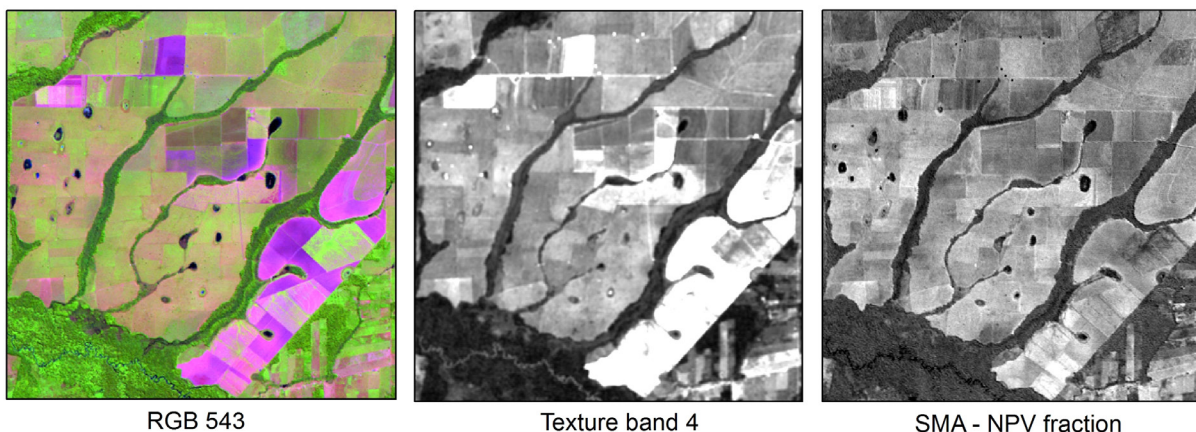
infrared bands to estimate texture images (mean algorithm). The texture and SMA images (Figure 3) enhanced presence of palm and other tree species within pasture lands, which helped to differentiate pasture lands from soybean lands.

Our results showed that soybean cropped areas substantially increased from 28,503 to 61,472 ha between 2000 and 2005, further 121,370 ha by 2010, and 147,812 ha by 2014 (Figure 4). The relatively small amount of soybean lands in 2000 was highly concentrated in Southern Rondônia, mostly in the municipality of Vilhena. In 2010, the distribution of soybean plantations was spread among different municipalities in Southern Rondônia: Vilhena (36,362 ha), Corumbiara (24,576 ha), Chupinguaia (15,836 ha), Cerejeiras (13,306 ha), and Pimenteiras do Oeste (13,228 ha). In 2014, the greatly expanded soybean area had spread further out to Central and Northern Rondônia (Figure 4). Our results are comparable to IBGE (2015) statistics, with each estimating a similar soybean planted area across Rondônia’s municipalities.

Soybean areas detected in 2000, 2005, 2010, and 2014 were overlapped with the deforestation dataset provided by the Amazonian Deforestation Satellite Monitoring Program (PRODES) (Table 3). This analysis showed that more than

**Table 2.** Accuracy assessment based on two reference SPOT images and the soybean cropped land classification.

		Reference image			Producer's accuracy	User's accuracy
		Soybean	Other land uses	Total		
Classified image	Soybean	30	0	30	77%	100%
	Other land uses	9	91	100	100%	91%
	Total	39	91	130		
Omission errors		23%	0%			
Commission errors		0%	9%			
Global accuracy		93%				



**Figure 4.** RGB 543 image, Texture image and SMA image. This figure is in color in the electronic version.

**Table 3.** Soybean cropped lands detected in 2000, 2005, 2010, and 2014 compared to the PRODES deforestation by 1997, 2000, 2005, 2010, and 2014.

		Soybean area (ha)			
		2000	2005	2010	2014
Period of deforestation	1997	22,582	40,795	84,392	106,939
	1997-2000	722	2,916	7,044	10,412
	2000-2005	-	15,126	25,460	23,479
	2005-2010	-	-	2,659	3,065
	2010-2014	-	-	-	3,745
Total*		28,503	61,742	121,370	147,812

\*The total soybean area that overlaps with forest class defined by the PRODES project (INPE, 2015). The remaining soybean areas by 2000, 2005, 2010, and 2014 were detected over non-forest class (*Cerrado* vegetation type) and, therefore, they were not here classified as deforestation.

95% of new soybean cropped lands were established on lands deforested nine or more years earlier, with 72% deforested by 1997, and 23% deforested between 1997 and 2005. Our results for Rondônia are consistent with previous research suggesting that cattle pastures are being replaced by soybean plantations in the Amazonian state of Mato Grosso (Pongratz *et al* 2006).

We observed that soybean fields tend to be concentrated in larger cropping fields in Southern Rondônia (Vilhena, Cerejeiras, Colorado do Oeste, Pimenteiras, Corumbiara, and Chupinguaia municipalities), a more consolidated region, while being more fragmented and smaller in Northern Rondônia (Porto Velho, Machadinho, Nova Mamoré, Cujubim and Buritis municipalities) as of 2014. In the recent northern soybean frontier, 58.3% of soybean fields were less than 100 ha, 41.7% were between 100 and 1000 ha, and none were larger than 1,000 ha. In comparison, 24.4% of soybean fields were less than 100 ha, 47.7% sized between 100 and 1000 ha, and 28% were larger than 1,000 ha within the older southern region.

## DISCUSSION

Despite the historic role of cattle raising as a major driver for the expansion of deforestation, soybean is rapidly becoming an important cause of land clearing in the Brazilian Amazon. The increase in soybean plantations is indirectly affecting the rates of deforestation in several Brazilian Amazon States (e.g., Rondônia, Pará, and Mato Grosso) because of the displacement of cattle ranching to areas still covered with primary forests (Barona *et al.* 2010). The relationship between soybean expansion and fresh deforestation is particularly important because farmers are moving from old colonization frontiers to remote areas in the Brazilian Amazon. In addition, instead of adopting the traditional cycle starting with annual crops and ending with pastures, these farmers may convert

forested lands directly into pastures (Brandão *et al.* 2005; Barona *et al.* 2010).

Barona *et al.* (2010) estimated that there is a high statistical correlation between deforestation and pastures and low correlation between deforestation and soybean expansion. Their results also indicated a high correlation between the decline of pasture quality and soybean expansion in the state of Mato Grosso. Assuming that a similar correlation might be applied to Rondônia, soybean plantations may have been replacing pastures after they reach a certain economic and environmental threshold. However, Barona and his collaborators could not conclude the extension to which deforestation in the Legal Amazon is related to pasture expansion or soybean expansion. Accordingly, Barona *et al.* (2010) and Morton *et al.* (2006) observed that while cattle raising is expanding into forest edges, mechanized agriculture (e.g. soybean plantations) is concentrating on old colonization areas in the Brazilian Amazon.

Moreover, pastures are more likely to be converted to soybean plantations when the profits of soy production increase and, consequently, cause the displacement of cattle raising to new colonization areas (Arima *et al.* 2011). Macedo *et al.* (2011) confirm that a wide range of forested areas were replaced by pasture and, subsequently, converted into soybean areas in the southern Brazilian Amazon. Macedo and her collaborators have estimated that between 2000 and 2005, 91% of pastures in the state of Mato Grosso were replaced by soybean plantations.

Our results lend support to the model of cattle raising displacement operations from older frontiers to new colonization areas occurring in Rondônia. Spatial patterns of land concentration and movement of soybean plantation over the years may indicate that small farmers and ranchers are increasingly being displaced from the older consolidated regions of southern Rondônia towards frontier areas of Northern Rondônia where standing forests are steadily being converted into pasture lands. Consequently, in the north, there has been a large increase in cattle herd numbers and deforestation rates (126,996 head year<sup>-1</sup> and 774,300 ha year<sup>-1</sup>, respectively) between 2002 and 2013 (IBGE 2015; INPE 2015). Meanwhile, cattle numbers and deforestation rates have increased only modestly (5,980 head year<sup>-1</sup> and 115,691 ha year<sup>-1</sup>, respectively) in the southern region during the same period of time (IBGE 2015; INPE 2015).

Recently, several authors have mentioned that the Soy Moratorium and other policies have contributed to declining deforestation rates in the Brazilian Amazon. The Soy Moratorium was signed by major soybean traders in an agreement not to purchase any soybeans grown on lands deforested after July 24<sup>th</sup>2006. Our research results suggest that many forested areas in Rondônia that were cleared

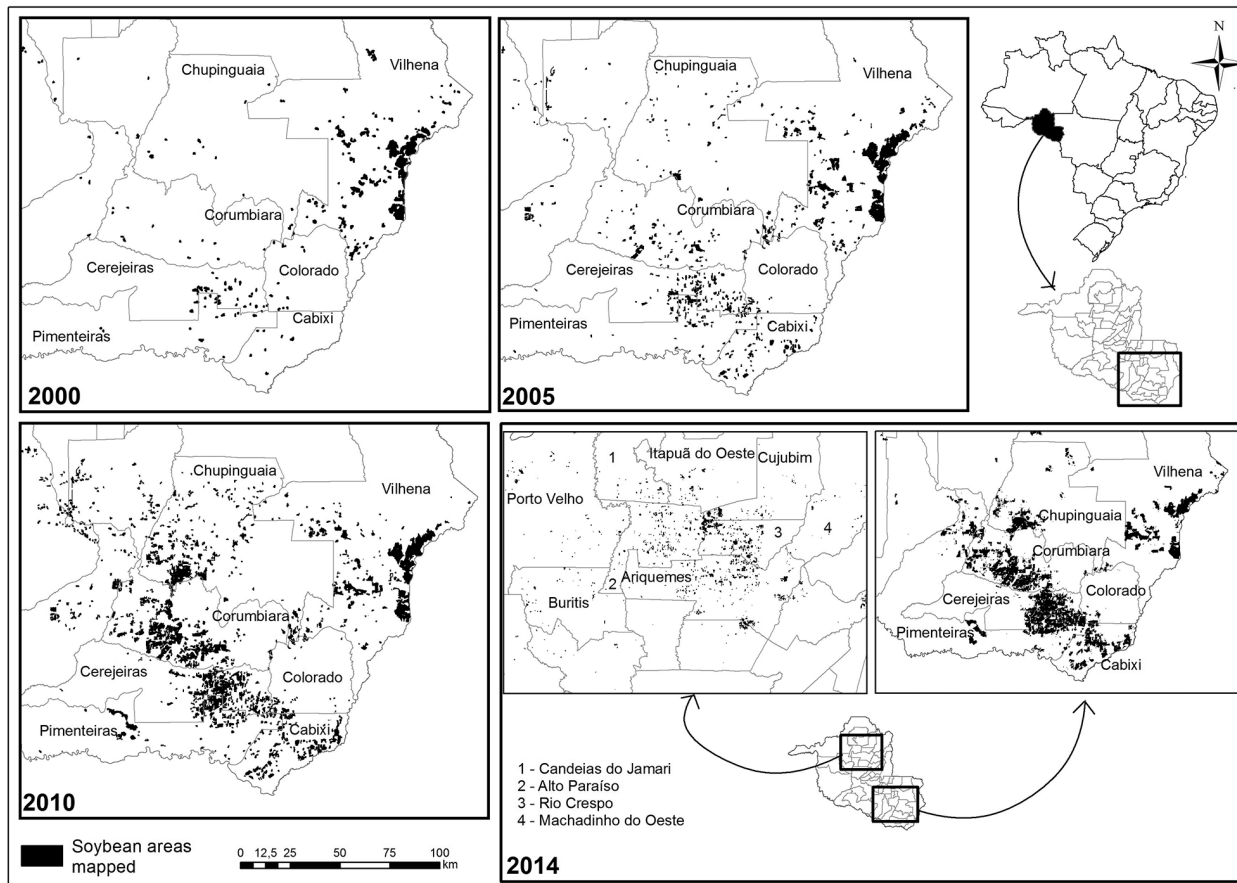


Figure 5. Soybean cropped lands in the state of Rondônia in 2000, 2005, 2010, and 2014.

and converted to pasture since 2006 are now economically viable for conversion to soybean cropping. Therefore, the effectiveness of the Soy Moratorium, as reported by Rudorff *et al.* (2011) and the hypothesis of the positive effects of the soy supply chain to reduce deforestation rates in the Amazon region, as suggested by Nepstad *et al.* (2014) and Gibbs *et al.* (2015) may only now be directly assessed for the case of Rondônia.

## CONCLUSIONS

This study presented an alternative approach to map soybean plantations in the Brazilian Amazon based on remotely sensed data and techniques and field work combined. Our soybean mapping showed a new increasing frontier of soybean cropped areas in Northern Rondônia in recent years. We concluded that soybeans plantations are possibly replacing old deforestation sites and moving those former land uses to new colonization of fresh deforestation areas. Moreover, soybean plantations can only settle in areas with 9 or more years after deforestation, in compliance with the dictates of the Soy Moratorium. Finally, we argue that, because soybean

cultivations are a lucrative and increasingly widespread land use, its impacts on the overall rates of deforestation have yet to be accurately quantified or appropriately considered by policymakers for the Brazilian Amazon.

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