



Monitoring burned area, its associated carbon emissions and forest fire risk Liana O. Anderson

liana.anderson@cemaden.gov.br

Cemaden:

Christopher Cunningan, Missae Yamamoto

INPE:

Egidio Arai, Celso Silva Junior, Luaê Andere, Marisa Fonseca, Natalia Salazar, Thais Rosan, Valdete Duarte, Yosio Shimabukuro, Luiz Aragão

Acre:

Instituto de Mudanças Climáticas (IMC), UFAC, SEMA

Roraima:

EMBRAPA

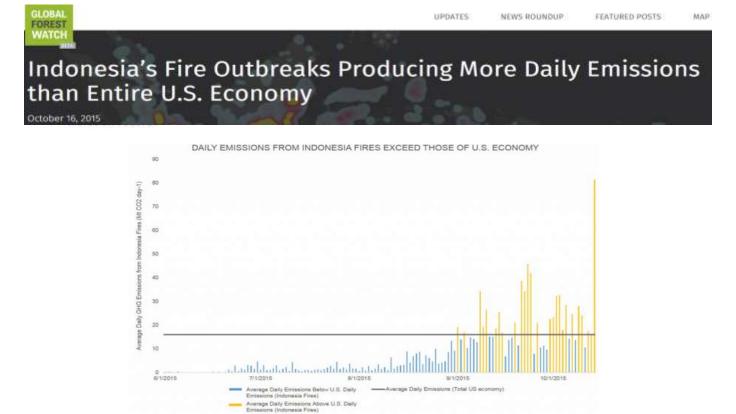
Maranhão:

UNICEUMA, UFMA, ICMBio

M Gloor (University of Leeds), S.Saatchi (JPL-NASA), Y. Malhi (University of Oxford)



La Red Latinoamericana de Teledetección e Incendios Forestales 17 Nov 2015



SOURCE: GLOBAL FIRE EMISSIONS DATABASE and CAIT

Biomass burning in South America emits on average 15 % of total global fire emissions

ORLD RESOURCES INSTITUTE





Monitoring burned area Carbon Emissions Fire risk

La Red Latinoamericana de Teledetección e Incendios Forestales 17 Nov 2015





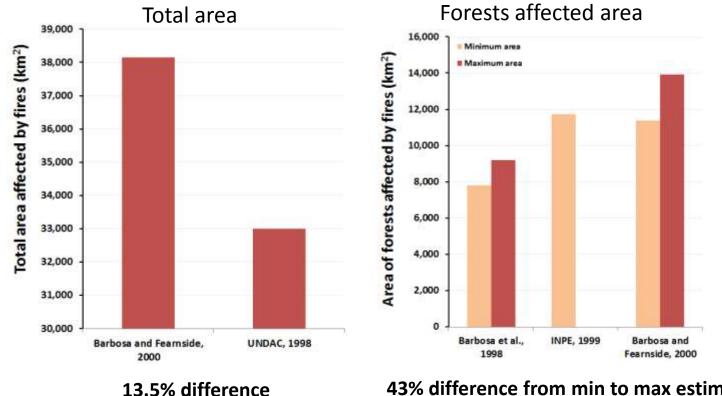
1. Monitoring burned area

La Red Latinoamericana de Teledetección e Incendios Forestales 17 Nov 2015

1. What we know about fire regime in SA?

We have some estimates of burned area with contrasting results.....

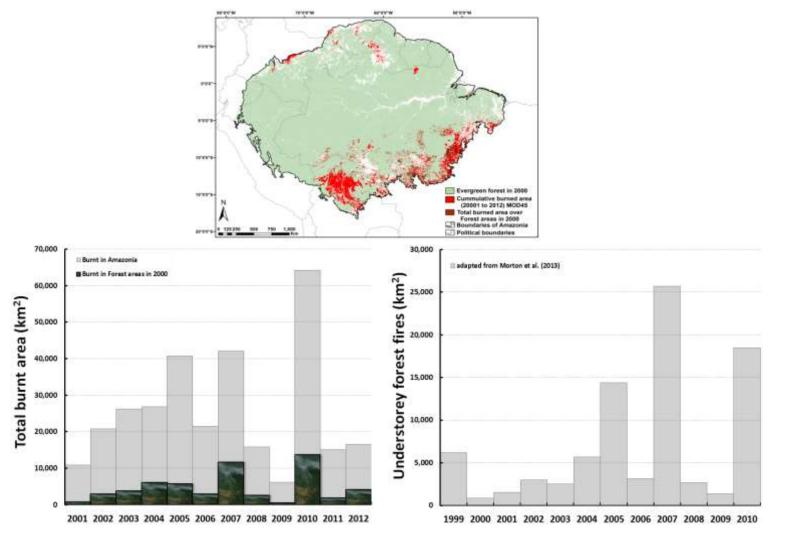
During the El Niño event in 1997-1998 in Roraima:



43% difference from min to max estimates

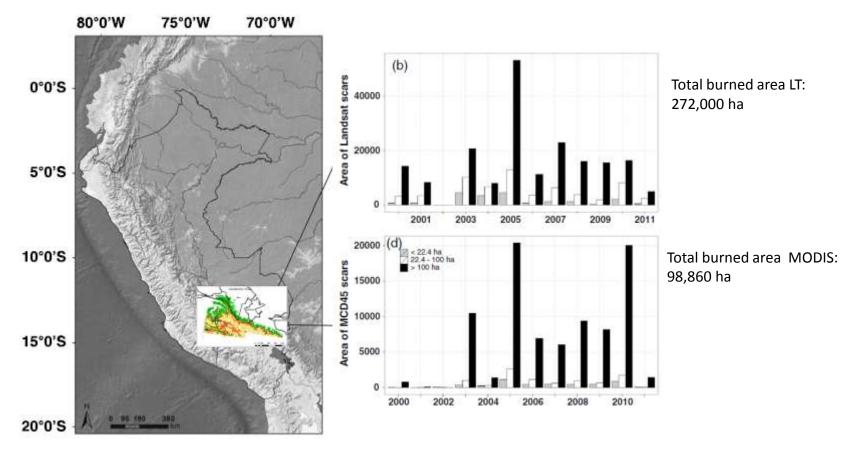
1. What we know about fire regime in SA?

And only MODIS-based products with a consistent temporal coverage (2000 onwards)



1. What we know about fire regime in SA?

And only MODIS-based products with a consistent temporal coverage (2000 onwards) AND We know that it underestimates the total burned area



FRI for the area is 37 years for grasslands, which is within the range reported for grasslands, and 65 years for forests



ACTA AMAZONICA

Detecção de cicatrizes de áreas queimadas baseada no modelo linear de mistura espectral e imagens índice de vegetação utilizando dados multitemporais do sensor MODIS/TERRA no estado do Mato Grosso, Amazônia brasileira

2005

Liana Oighenstein ANDERSON¹, Luiz Eduardo Oliveira e Cruz de ARAGÃO¹, André de LIMA¹, Yosio Edemir SHIMABUKURO¹

RESUMO



Applied Geography Volume 34, May 2012, Pages 239-246



Land use and land cover changes determine the spatial relationship between fire and deforestation in the Brazilian Amazon

André Lima[®] • [©], Thiago Sanna Freire Silva[®] [©], Luiz Eduardo Oliveira e Cruz de Aragão[®] [©], Ramon Morais de Feitas[®] [©], Marcos Adami[®] [©], Antônio Roberto Formaggio[®] [©], Yosio Edemir Shimabukuro[®] [©]

Show more

doi:10.1016/j.apgeog.2011.10.013

Got rights and content

Abstract

An increased frequency of droughts is predicted for the Amazon rainforest in the 21st century, which, combined with deforestation, could exacerbate fire occurrence in the Assessment of Deforestation in Near Real Time Over the Brazilian Amazon Using Multitemporal Fraction Images Derived From Terra MODIS

2005

Liana O. Anderson, Yosio E. Shimabukuro, Ruth S. Defries, and Douglas Morton

Abstract—We present a methodology for rapidly assessing deforestation over the Annaon region method for policy intervention. We use soil fraction images generated from Moderate Resolution Imaging Spectroradiometer (MODIS) data at 250-m spatial resolution. Results showed reasonable agreement with higher resolution Landsait data ($r^2 = 0.73$) for our study area. MODIS data are promising for near real-time deforestation monitoring, previously not practical with Landsait data.

HER GRONCHINCE AND REMOTE SENSING LETTERS. VOL. 2, NO. 3, JULY 2005

Index Ternu—Amazon region, deforestation, fraction images, Moderate Resolution Imaging Spectroradiometer (MODIS) sensor. pixel value is a mixture of spectral response from different targets within each pixel, and it is more evident in medium or low spatial resolution imoges. Several techniques, such as modeling and empirical estimations, have been applied to depict subpixel heterogeneity in land cover from remotely sensed data [3], [4].

Fraction images derived from different remote sensing data have been used for monitoring reforested areas [5], deforestation [6], land cover change [7], and vegetation classification [8]. These images derived from a linear spectral mixing model constitute synthetic bands containing information about proportions.

International Journal of Remote Sensing

Volume 30, Issue 6, 2009



Article

Fraction images derived from Terra Modis data for mapping burnt areas in Brazilian Amazonia

2009

DOI: 10.1080/01431160802509058

Y. E. Shimabukuro^{a*}, V. Duarte^a, E. Arai^a, R. M. Freitas^a, A. Lima^a, D. M. Valeriano^a, I. F. Brown^b & M. L. R. Maldonado^b pages 1537-1546

Remote Sensing 2014, 6(9), 8002-8025; doi:10.3390/rs6098002



Doen Acce

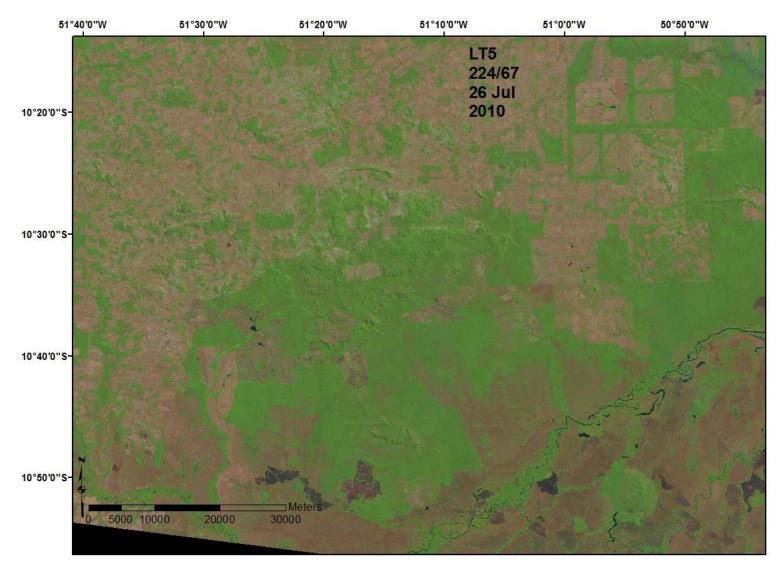
Analysis and Assessment of the Spatial and Temporal Distribution

of Burned Areas in the Amazon Forest

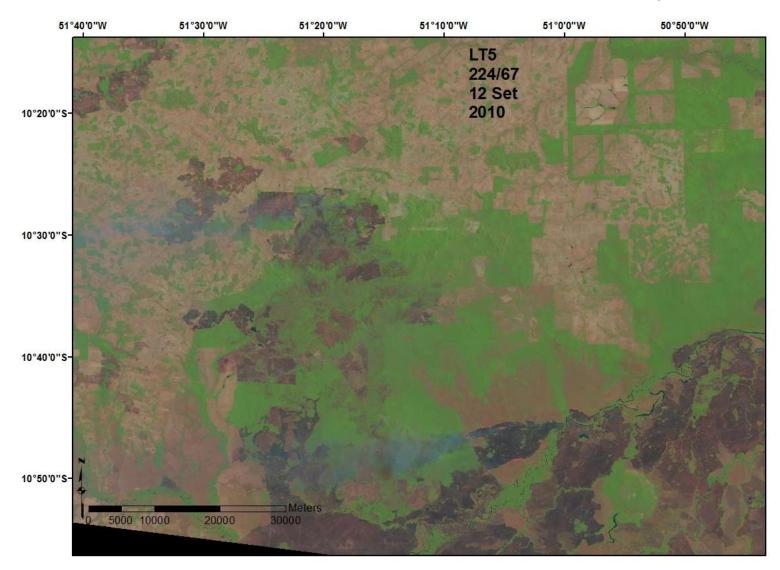
Francielle da Silva Cardozo^{1,*}, Gabriel Pereira², Yosio Edemir Shimabukuro¹ and Elisabete Caria Moraes¹

Remote Sensing Division (DSR), National Institute for Space Research (INPE), Av. dos Astronautas, 1758, Jd. Granja, CEP: 12227-010, Sao Jose dos Campos, Brazil; E-Mails: yosio@dsr.inpe.br

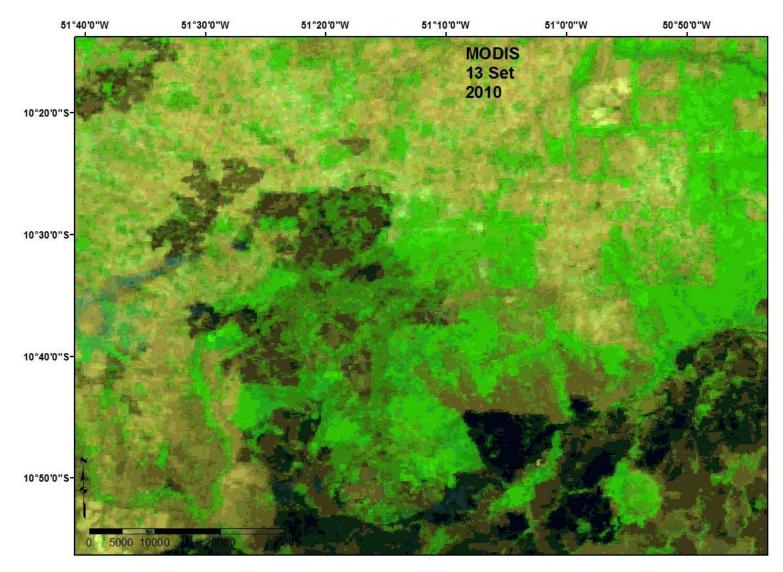
Landsat 5 – TM Surface Reflectance RGB 543, 26th Jul 2010



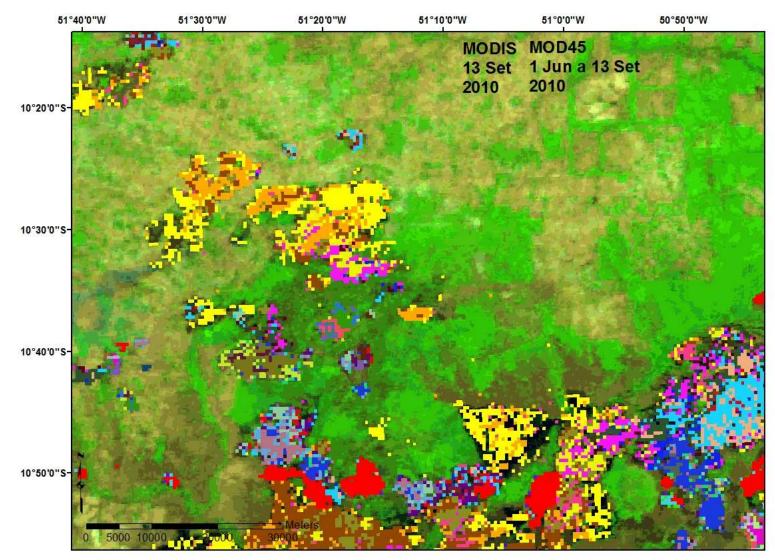
Landsat 5 – TM Surface Reflectance RGB 543, 12th Sep 2010



Product MOD09 – Surface Reflectance RGB 126, 13th Set 2010

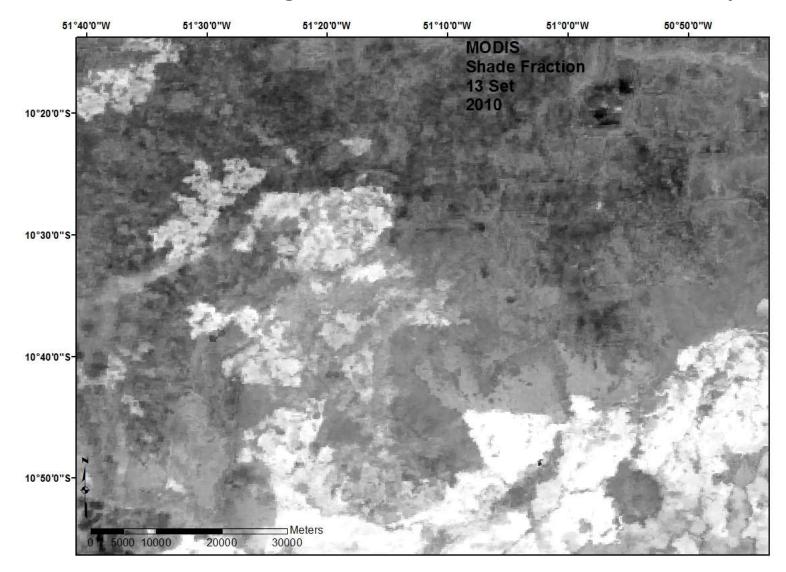


Product MOD45 - MODIS Burned Area Product

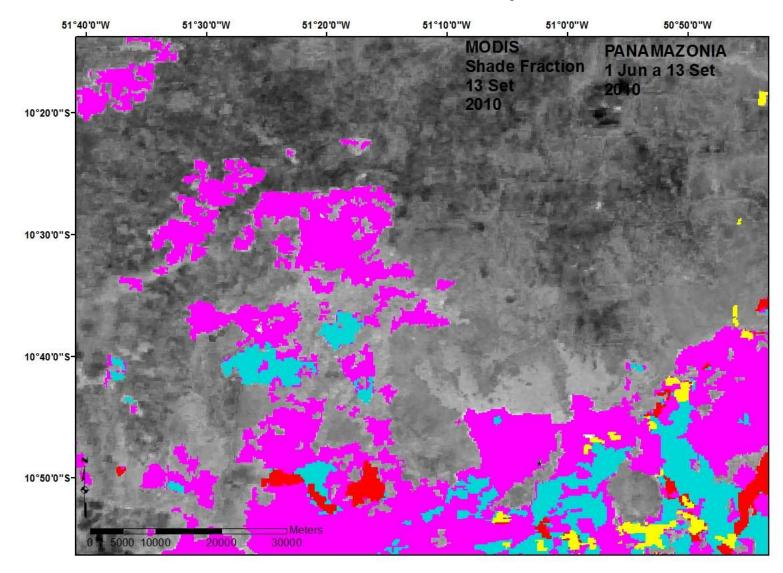


Subestimativa de áreas afetadas pelo fogo

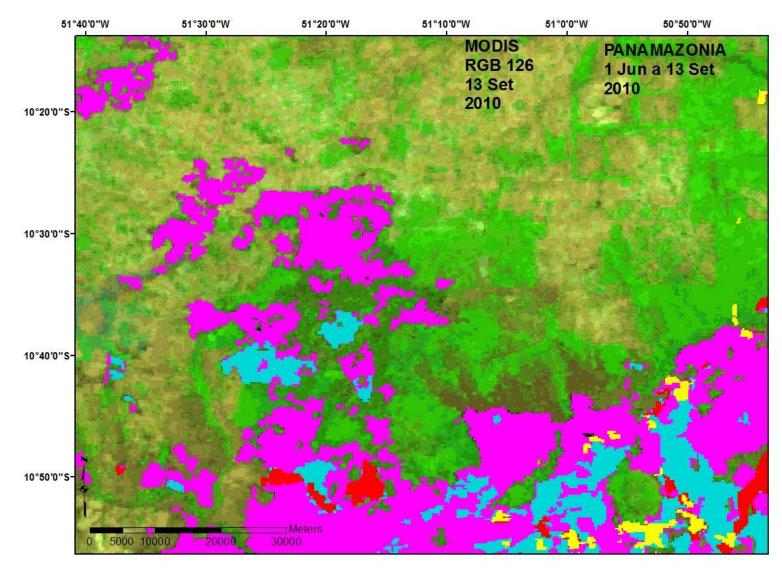
Shade fraction image based on the Product MOD09 – 13 Sept 2010



Burned area classification – 13 Sept 2010

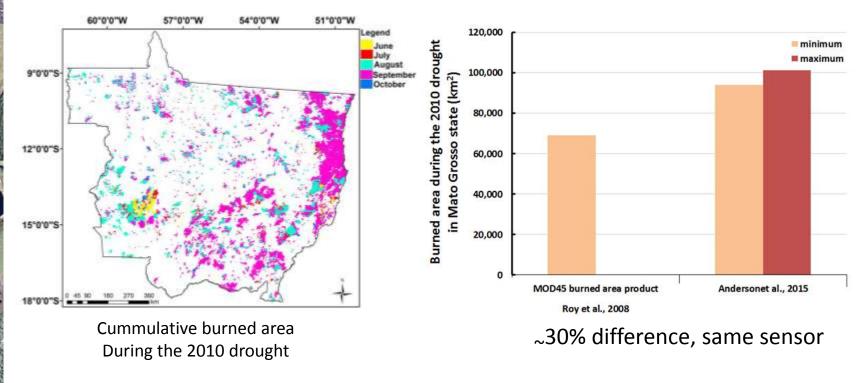


Cummulative Burned area based on our method – 13 Sept 2010



1. What we know about fire regime in SA?

And only MODIS-based products with a consistent temporal coverage (2000 onwards) AND We know that it underestimates the total burned area



31% to 33% of the total pristine vegetation remaining in Mato Grosso burnt in 2010.

Anderson et al 2015, Gblobal Biogeochm. Cycles. Anderson et al 2015 under review RS

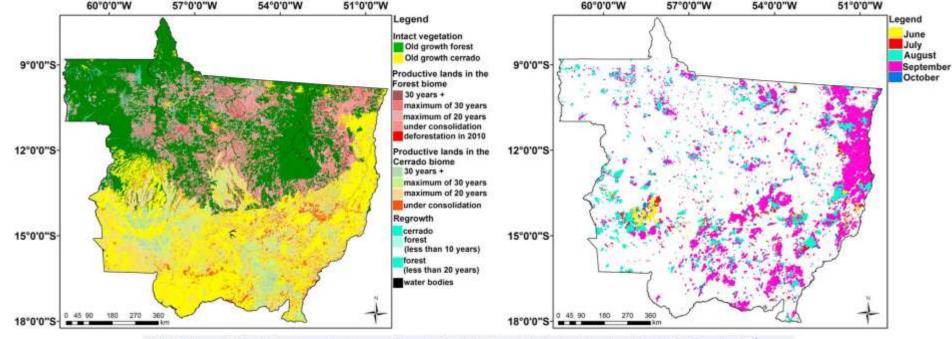


Table 2. Biomass of Affected Area. Biomass Loss, Gross, and Committed C Emissions Due to Fires Per Land Cover Type During the Dry Season in 2010⁸

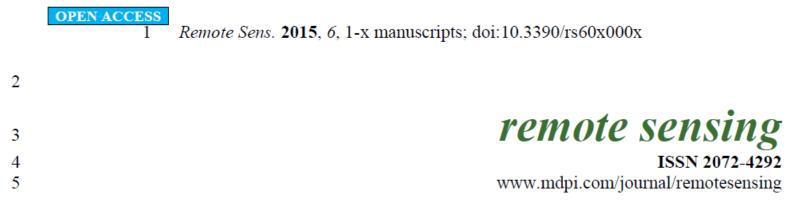
	Biomass (Mg ha ⁻¹) of the Affected Areas Mean (±Error)		Gross C Emission in 2010 (Tg) (±Total Error)	Committed C Emission (Tg) (±Total Error)	Total Carbon Loss (Tg) (±Total Error)	% of Carbon Los
Intact vegetation						
Old-growth forest	176.3 (±34.63)	51.63 (±19.6)	5.05 (±1.92)	27.2 (±10.4)	32.3 (±12.28)	37.87
Old-growth cerrado	40.23 (±33.43)	11.71(±4.79)	34.1 (±13.9)		34.1 (±13.94)	39.98
Productive lands in the forest biome						
Permanent productive for 30 years +	53.69 (±33.27)	15.65 (±6.40)	0.46 (±0.19)	1.00	0.46 (±0.19)	0.54
Permanent productive for maximum of 30 years	68.42 (±33.67)	19.95 (±8.16)	0.99 (±0.40)	1280	0.99 (±0.40)	1.16
Permanent productive for maximum of 20 years	105 (±34.62)	30.61 (±12.5)	5.13 (±2.10)		5.13 (±2.09)	6.01
Under consolidation (productive for 10 years or less)	105 (±34.76)	17.29 (±7.07)	5.02 (±2.05)	1.00	5.02 (±2.053)	5.89
Deforestation in 2010	208.9 (±33.24)	60.91 (±23.0)	4.78×10^{-3}	3.01×10^{-2}	9.56×10^{-3}	0.01
			$(\pm 1.81 \times 10^{-4})$	$(\pm 2.71 \times 10^{-2})$	$(\pm 3.61 \times 10^{-3})$	
Productive lands in the cerrado biome			And a second sec	and a second		
Permanent productive for 30 years +	27.17 (±32.09)	7.92 (±3.24)	1.00 (±0.41)	1.141	1.00 (±0.41)	1.17
Permanent productive for maximum of 30 years	33.5 (±32.39)	9.76 (±3.99)	1.22 (±0.49)		1.22 (±0.49)	1.43
Permanent productive for maximum of 20 years	31.14 (±32.22)	9.08 (±3.71)	1.99 (±0.81)		1.99 (±0.81)	2.33
Under consolidation (productive for 10 years or less)	31.14 (±32.61)	9.29 (±3.80)	9.75×10^{-2} (±1.22 × 10 ⁻²)		9.75 × 10 ⁻² (±0.21)	0.11
Reprovth			State of the second sec		10000 M	

31% to 33% of the total pristine vegetation remaining in Mato Grosso burnt in 2010.

The second se	1000 CONTRACTOR 1000 CONTRACTOR	Marcine and Constantial	(±3.87×10)	(±2.09 × 10)	(±2,47×10)	method
Forest regrowth (less than 10 years)	148.8 (±34.82)	49.74 (±18.9)	0.30 (±0.11)	1.64 (±0.62)	1.95 (±0.74)	2.29
Deforestation in 2010 on less than 10 years regrowth	131.2 (±34.76)	85.27 (±0.0)	0.45 (±0.0)	0.45 (±0.0)	0.91 (±0.0)	1.07
Total			56.1 (±22.5)	29.4 (±10.0)	85.3 (±33.2)	100

⁸The error associated with the biomass of affected areas was derived from the AGB error map from Saatchi et al. [2011]. The errors associated with biomass loss, carbon loss, and emissions were derived from the Monte Carlo simulations.

Determining confidence interval and Accuracy



⁶ Article

Development of a Point-based method for Map Validation and
Confidence Interval Estimation: A case study of Burned Areas

- 9 **in Amazonia**
- 10

11 Liana Oighenstein Anderson ^{1,2,3†*}, David Cheek ^{2,†}, Luiz E.O.C. Aragão ^{3,4}, Luae Andere ³,

12 Brenda Duarte ³, Natália Salazar ³, André Lima ³, Valdete Duarte ³ and Egidio Arai ³

13 ¹ National Center for Monitoring and Early Warning of Natural Disasters – Cemaden. Parque

14 Tecnológico de São José dos Campos, Estrada Doutor Altino Bondensan, 500, São José dos

15 Campos - São Paulo, Brasil, 12247-016. E-Mail: liana.anderson@cemaden.gov.br

² Environmental Change Institute, University of Oxford, OXford, OX1 3QY, UK 7 E-Mail:

Determining confidence interval and Accuracy

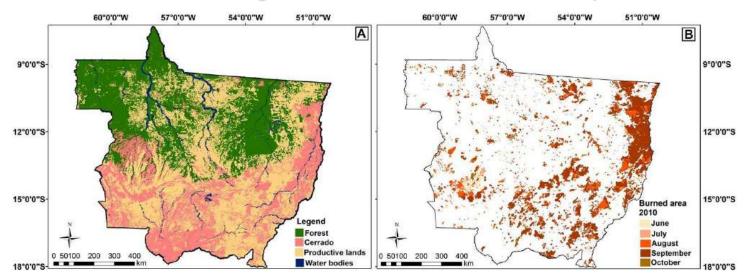
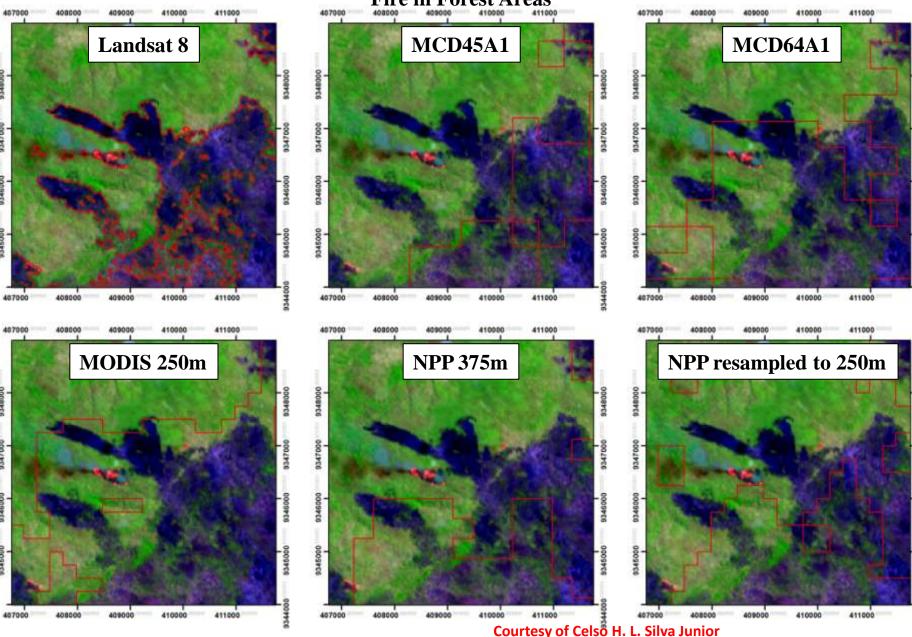


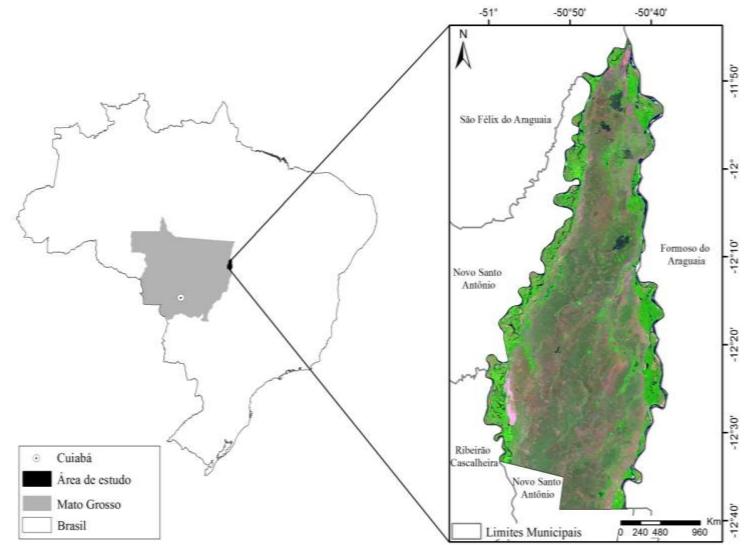
Figure 1: (a) Land cover classes in 2010 for Mato Grosso state. The forest class encompasses forests and forests regrowth, while non-forests are all the other land cover types; (b) Mapped burned areas from June to mid-October 2010.

By considering the area uncertainty provided by the validation scheme, the old growth forest area burned ranged from 13,678 km² to 13,929 km². The extent of the 2010 forest fires in Mato Grosso are close to southern Amazonian intact forests that were affected by fires in 2010 for the first time (13,570 km²), or 73% to 75% of all understory forest fires mapped for 2010 in southern Amazonia (18,499 km²) [55].

Non-forests burned areas were estimated to have affected from 78,089 km² to 83,107 km², an area approximately 6 times larger than the estimates of the 2010 burned forests. It has been

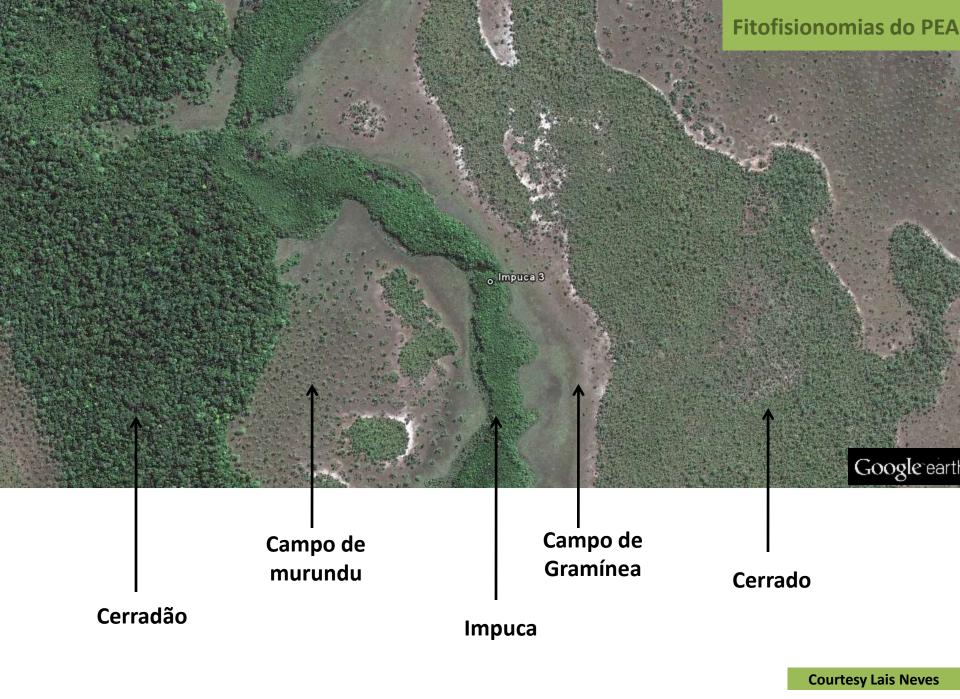


Área de estudo



Parque Estadual do Araguaia no Mato Grosso. Composição colorida (RGB) 5, 4, 3 da imagem do Landsat-7/ETM+, da órbita/pontos: 233/68 e 69 de 18/07/2001.

Courtesy Lais Neves

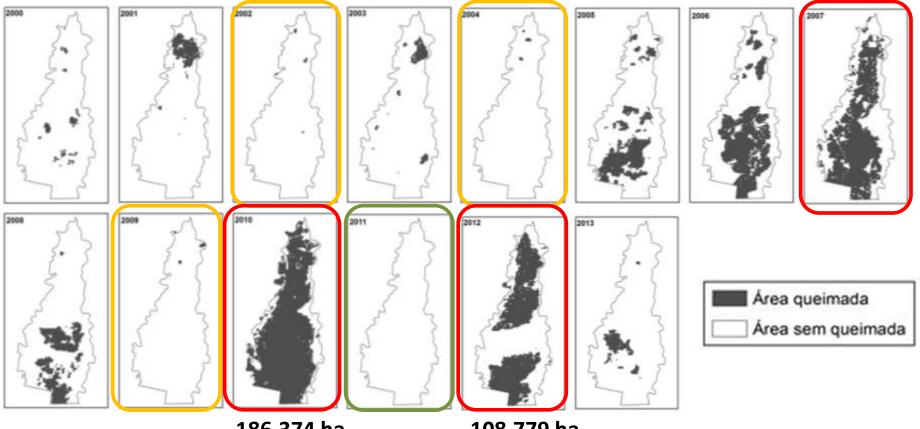


Prof^a. Beatriz S. Marimon

Resultados

Padrão temporal dos incêndios

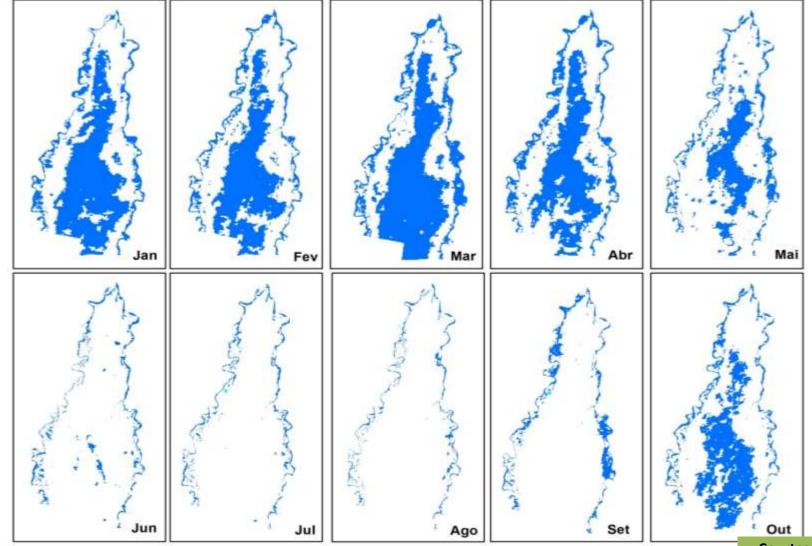
129.540 ha



108.779 ha

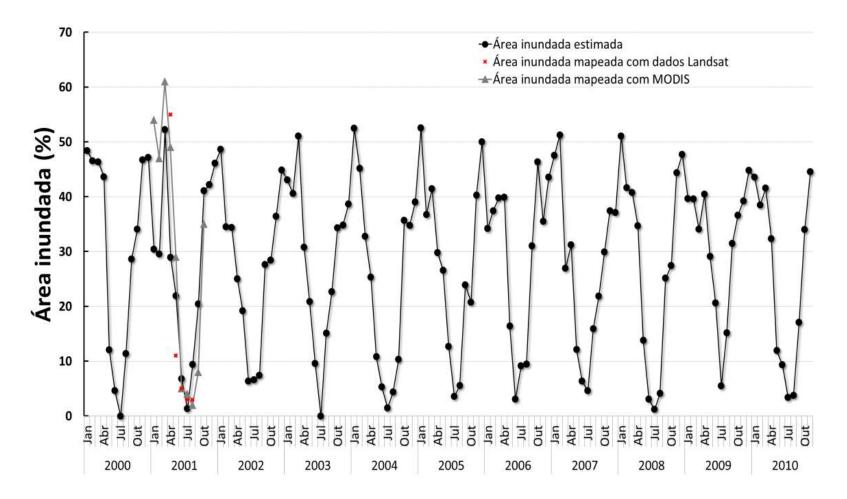
Resultados

Dinâmica de Inundação



Courtesy Lais Neves

Extrapolação da área inundada



Série histórica de inundação do Parque Estadual do Araguaia entre 2000 e 2010.

Courtesy Lais Neves





2. Carbon emission from fires

La Red Latinoamericana de Teledetección e Incendios Forestales 17 Nov 2015

Brazilian National Plan on Climate Change Decree number 7.390/2010

PROJECTION TO THE YEAR 2020: of 0,87 Pg C TARGET AIMED FOR 2020: 0,53 Pg C to 0,55 Pg C, from which 0,25 Pg C is the projected emission from deforestation in the Amazon.

Our aircraft measurements:

During the 2011 anomalous <u>wet year</u>: fires were responsible for 0.30 ± 0.10 Pg C (higher than the projected reductions aimed from deforestation)

During the extreme <u>dry year</u> of 2010: fires were responsible 0.51±0.12 Pg C, an emission close to the total reduction planed by the Brazilian government for the year 2020.

[Gatti, ...Anderson et al., Nature 2014]

To understand the role of fires on C emissions and effectively support and drive the <u>development of mitigation strategies</u>, <u>including fire prevention and management</u>, it is necessary to **break down emissions** into natural forest and savannah fires, land management fires and deforestation-related fires.

Moreover, it is important to quantify the role of the different fire types, as only <u>wildfires in denser vegetation</u> or fires associated with deforestation represent a long-term net source of CO_2 to the atmosphere.

In summary, our model of gross emission follows:

$$F = \lambda_{(Idcover)} \cdot B_{i(x,y)} \cdot (1 - \alpha_{(x,y)}) dA_{(x,y)},$$

$$F' = \lambda'_{(ldcover)} \cdot B'_{i(x,y)} \cdot (1 - \alpha'_{(x,y)}) dA'_{(x,y)} (Monte Carlo)$$

where:

F (ton yr⁻¹) is carbon the gross emission (immediate flux to the atmosphere).

 $\lambda_{(Idcover)}$ is the 'decay or release constant (yr⁻¹) specific to each land cover type (ha).

 $\mathbf{B}_{i(x,y)}$ is the pre-burn biomass density (Mg ha⁻¹) for the pixel at location (x,y)

 α is the slope of the Equation 1 (α = 0.7084)

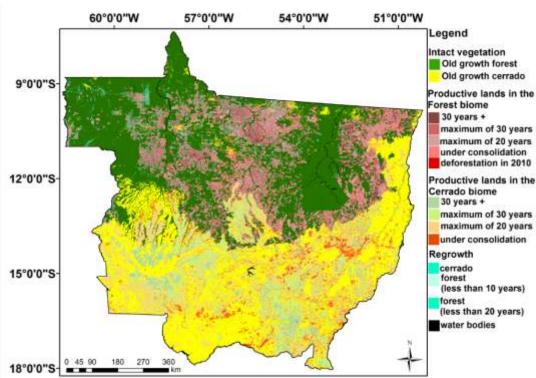
 $dA_{(x,y)}$ is the burned area (in ha) at pixel with location (x,y).

In summary, our model of gross emission follows:

$$F = \lambda_{(Idcover)} \cdot B_{i(x,y)} \cdot (1 - \alpha_{(x,y)}) dA_{(x,y)},$$

 $\lambda_{(Idcover)}$ is the 'decay or release constant (yr⁻¹) specific to each land cover type (ha).

Identification of land cover type has an accuracy of 80% (i.e. in 80% of the cases the land classification is correct, with a confidence interval from 74% to 87%) > Varying randomly this parameter uniformly distributed.



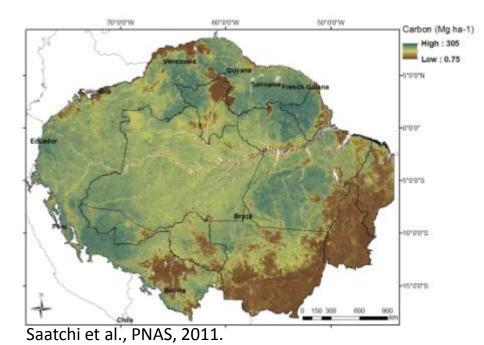
In summary, our model of gross emission follows:

$$F = \lambda_{(Idcover)} \cdot B_{i(x,y)} \cdot (1 - \alpha_{(x,y)}) dA_{(x,y)},$$

where:

 $\mathbf{B}_{i(x,y)}$ is the pre-burn biomass density (Mg ha⁻¹) for the pixel at location (x,y) provided by Saatchi et al. [2011].

Forest and non-forest pixels > varying randomly this parameter with an approximately normally distributed.



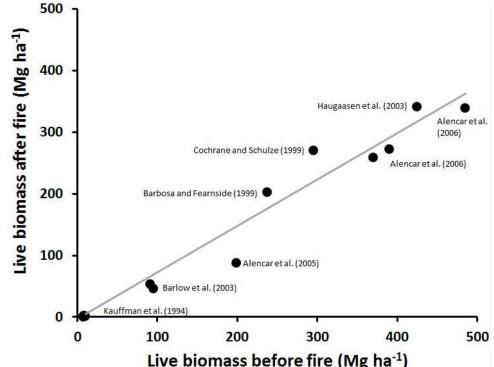
Anderson et al., 2015 Global Biogeoch. Cycles

In summary, our model of gross emission follows:

$$F = \lambda_{(\text{Idcover})} \cdot B_{i(x,y)} \cdot (1 - \alpha_{(x,y)}) dA_{(x,y)},$$

where:

 α is the slope of the Equation $B_f = 0.7084^*B_i$ (B = biomass, f = final, i = initial) $\alpha \pm 0.034$ (Mg Mg⁻¹) slope and its uncertainty is estimated using a linear regression which takes into account both x and y errors [York et al., 2004, Unified equations for the slope, intercept, and standard errors of the best straight line. Am. J. Phys. 72 (3)]



Anderson et al., 2015 Global Biogeoch. Cycles

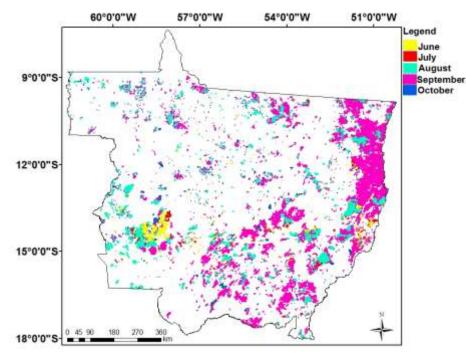
In summary, our model of gross emission follows:

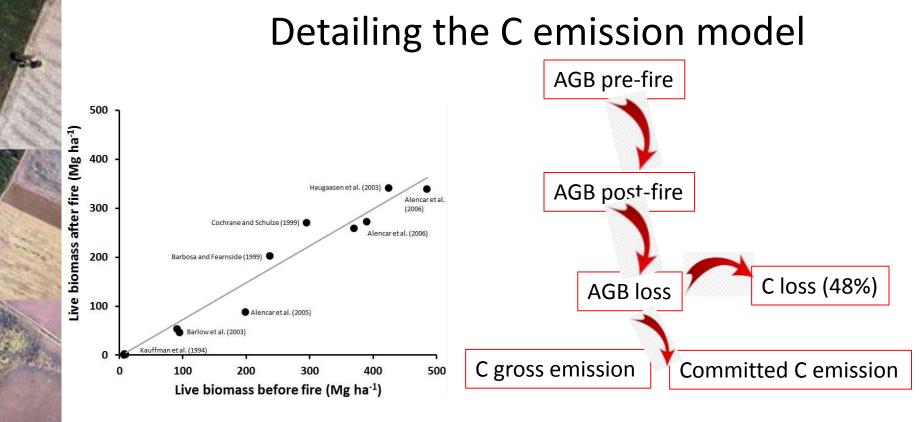
$$F = \lambda_{(Idcover)} \cdot B_{i(x,y)} \cdot (1 - \alpha_{(x,y)}) dA_{(x,y)},$$

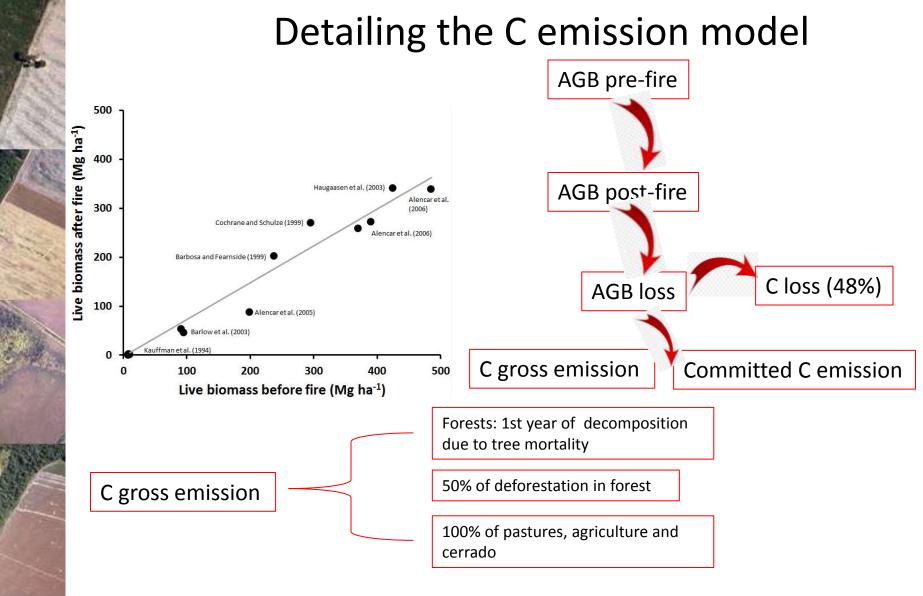
where:

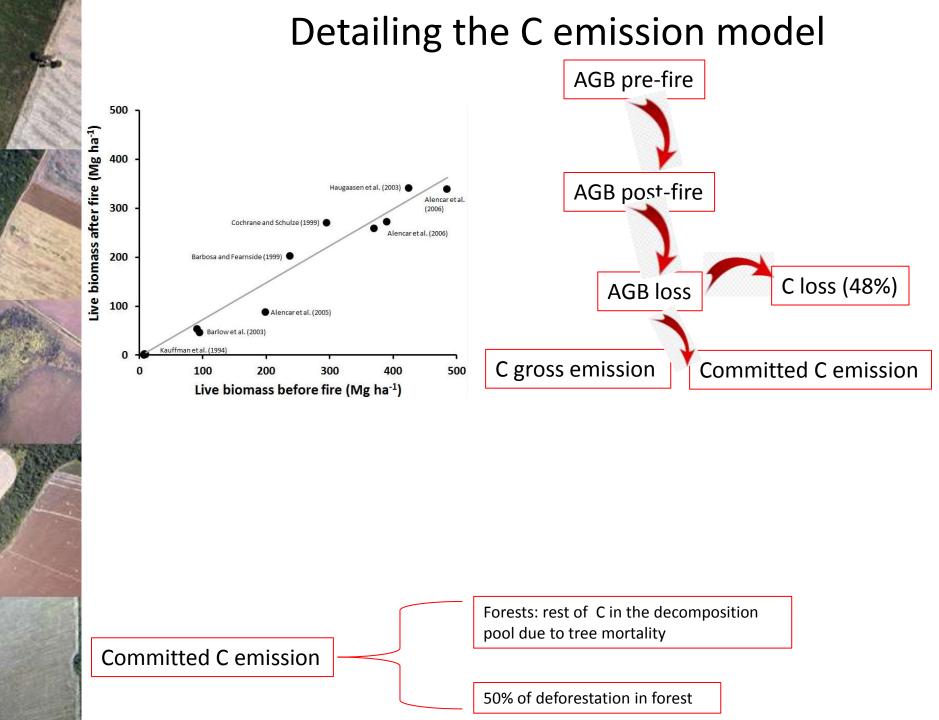
 $dA_{(x,y)}$ is the burned area (in ha) at pixel with location (x,y).

An accuracy of 99.2% (<u>confidence interval</u> of 97.67% to 99.48%) for forests and an accuracy of 96.93% (<u>confidence interval</u> of 93.76% to 98.92%) for the non-forest classes > Varying randomly this parameter uniformly distributed.

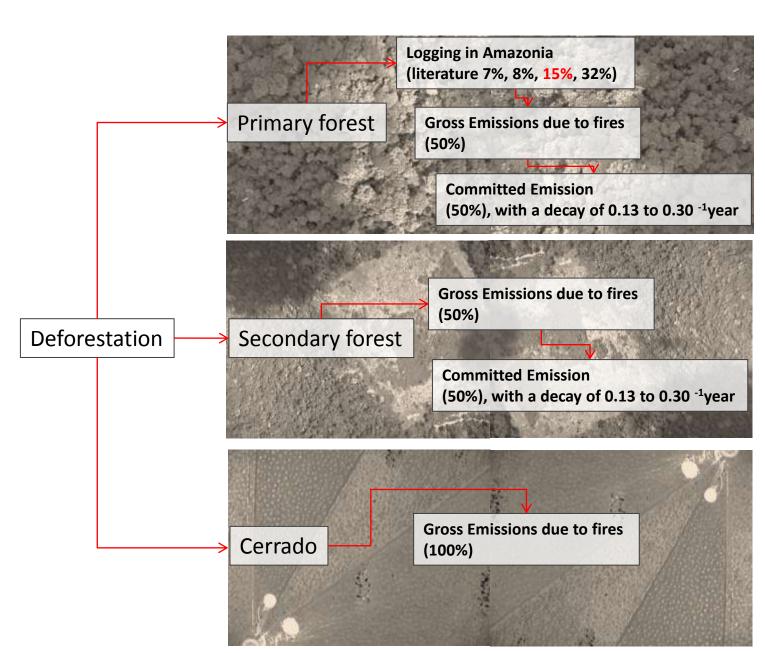








Detailing the C emission model



2. Quantifying the associated C emissions

	Biomass (Mg ha ⁻¹) of the affected areas Mean (± error)	Biomass loss due to fires (Mg ha ⁻¹) Mean (± error)	Gross C emission in 2010 (Tg) (±Total error)		Total Carbon loss (Tg) (±Total error)	% of Carbon loss
Intact vegetation			-			
Old growth Forest	176.3 (±34.63)	51.63 (±19.6)	5.05 (±1.92)	27.2 (±10.4)	32.3 (±12.28)	37.87
Old growth Cerrado	40.23 (±33.43)	11.71(±4.79)	34.1 (±13.9)		34.1 (±13.94)	39.98
Productive lands in the Forest biome						
Permanent productive for 30 years +	53.69 (±33.27)	15.65 (±6.40)	0.46 (±0.19)		0.46 (±0.19)	0.54
Permanent productive for maximum of 30 years	68.42 (±33.67)	19.95 (±8.16)	0.99 (±0.40)		0.99 (±0.40)	1.16
Permanent productive for maximum of 20 years	105 (±34.62)	30.61 (±12.5)	5.13 (±2.10)		5.13 (±2.09)	6.01
Under consolidation (productive for 10 years or less)	105 (±34.76)	17.29 (±7.07)	5.02 (±2.05)		5.02 (±2.05 ³)	5.89
Deforestation in 2010	208.9 (±33.24)	60.91 (±23.0)	4.78x10 ⁻³ (±1.81x10 ⁻⁴)	$3.01 \times 10^{-2} (\pm 2.71 \times 10^{-2})$	9.56x10 ⁻³ ($\pm 3.61x10^{-3}$)	0.01
Productive lands in the Cerrado biome				•		
Permanent productive for 30 years +	27.17 (±32.09)	7.92 (±3.24)	1.00 (±0.41)		1.00 (±0.41)	1.17
Permanent productive for maximum of 30 years	33.5 (±32.39)	9.76 (±3.99)	1.22 (±0.49)		1.22 (±0.49)	1.43
Permanent productive for maximum of 20 years	31.14 (±32.22)	9.08 (±3.71)	1.99 (±0.81)		1.99 (±0.81)	2.33
Under consolidation (productive for 10 years or less)	31.14 (±32.61)	9.29 (±3.80)	9.75x10 ⁻² (±1.22x10 ⁻²))	9.75 x10 ⁻² (±0.21)	0.11
Regrowth						
Cerrado	37.51 (±32.55)	9.42 (±3.01)	9.21x10 ⁻² (±3.76x10 ⁻²)	l	9.21x10 ⁻² (±3.76x10 ⁻²)	0.11
Forest regrowth (less than 20 years)	208.1 (±34.11)	43.38 (±16.4)	1.03x10 ⁻³ (±3.87x10 ⁻⁴)	5.53x10 ⁻³ (±2.09 x10 ⁻³)	$6.56 \times 10^{-3} (\pm 2.47 \times 10^{-3})$	0.01
Forest regrowth (less than 10 years)	148.8 (±34.82)	49.74 (±18.9)	0.30 (±0.11)	1.64 (±0.62)	1.95 (±0.74)	2.29
Deforestation in 2010 on less than 10 years regrowth	131.2 (±34.76)	85.27 (±0.0)	0.45 (±0.0)	0.45 (±0.0)	0.91 (±0.0)	1.07
Total			56.1 (±22.5)	29.4 (±10.0)	85.3 (±33.2)	100





La Red Latinoamericana de Teledetección e Incendios Forestales 17 Nov 2015

Overview

Brazil

Vast Amazon wildfire destroys forest in Brazil and threatens uncontacted tribe



The blaze, which has burned for two months on indigenous land and spread across 100km at its peak, is suspected to have been started by illegal loggers

Jonathan Watts in Rio de Janeiro

Friday 30 October 2015 10.30 GMT

Informativo CENAD nº 560 - 14/10/2015 - Tarde

AMAZONAS

O Governo do Amazonas decretou onten (13/10) Estado de Emergência em 12 cidades por causa do número de incêndos. A decisão foi publicada no Dário Oficial do Estado (DOE). A vigência do decreto será de 90 días e abrange as cidades de: Manaus, Autazes, Caapiranga, Cansiro, Careiro da Várzea, Iranduba, Itacoatian, Manacapuru, Manaquin, Novo Airão, Presidente Figueiredo e Ro Peto da Eva.

AMAPA

A defesa civil Estadual informou a ocorrência de um incêndio que destrói uma área de reflorestamento na Vila Operária, próximo ao município de Serra do Navio. A grande quantidade de fumaça está afetando a população. Até o momento não há registro de desabrigados ou desalojados. Forte: CEDC/AP

RIO GRANDE DO SUL

Atualização dos dados 60 municípios atingidos 50.168 pessoas atingidas 5.795 pessoas desalojadas 3.475 pessoas em abrigos 10.212 residências atingidas

Taquari: 08 familias em abrigos. Estrela: 23 famílias em abrigos Rio Pardo: 30 famílias desabrigadas e 150 desafojadas.

Fonte: CEDEC/RS

PARANÁ e SANTA CATARINA: não houve alteração dos dados.

Atenciosamente,

Monitoramento e Operações / CENAD / SEDEC / MI 0800 644 0199 (Plantão 24h)



ERASIL Acesso à informação

ir pana o contexido 🚺 ir pana o reariu 🛃 ir pana a bassa 🛐 ir pana o cotopé 🚺

Integração Nacional



Target variable

Fire Risk (FR)

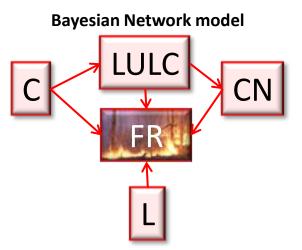
Context variables: exhibit any kind of relationship with the target variable and may exhibit relationships among themselves

Climatic variables (C) Land use and land

cover variables (LULC)

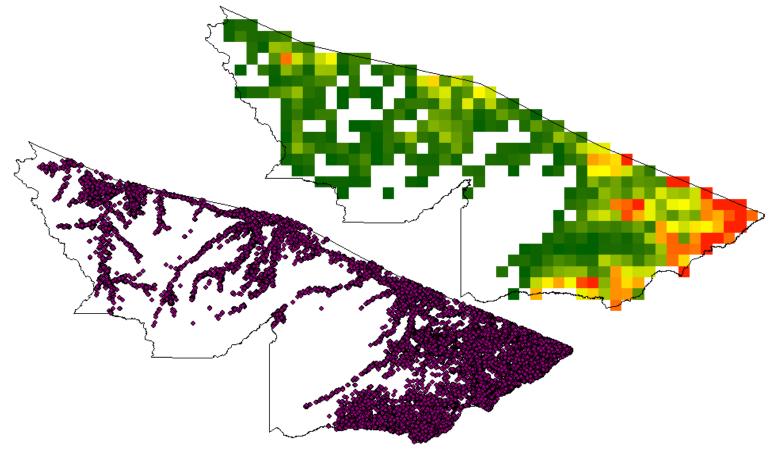
Census data (CN)

Local information (L)



Simplified Directed Acyclic Graph – DAG of my Fire Risk Model







Pilot project: Acre State

B = 0,05H - 0,1(T - 27)

B = índice de AngstronH = umidade relativa do ar em %

T = temperatura do ar em °C

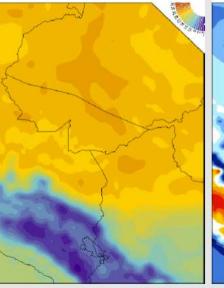
Dados: Sistema Regional de Modelagem Atmosférica BRAMS 5x5 Km

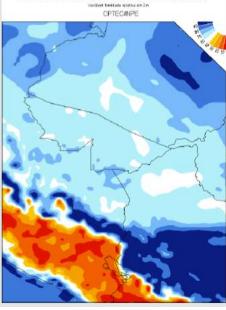
Temperatura

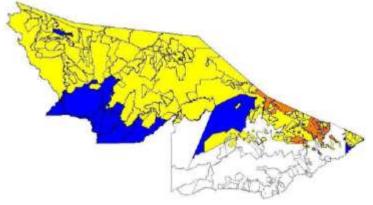
Umidade Relativa

saita-feisi) Vääda para: 88/7/2015, 00 UTC (Gaasta-feisi)

BRAMS 05 km Anabies Intel allisants emiliation de la Colombia de L







Processamento TerraMA²

NÍVEL DE ALERTA

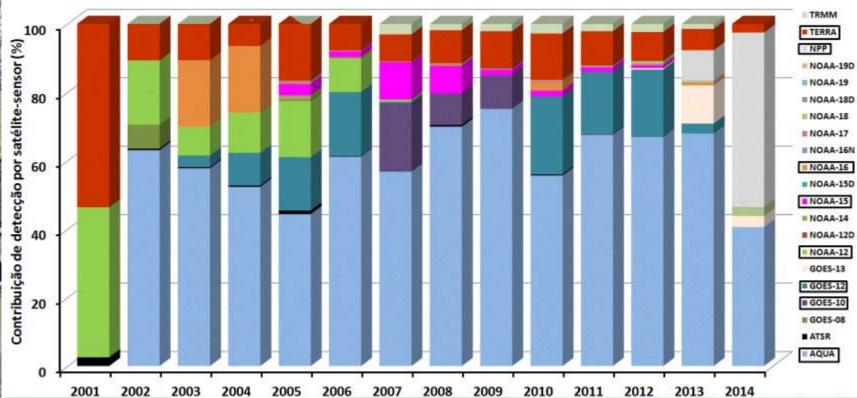
VALOR RISCO	ALERTA
3	Observação
2,5	Atenção
2	Alerta
1,5	Alerta Máximo

RESULTADOS

Envio de notificação por e-mail; Mapas de Risco; Acesso informação internet

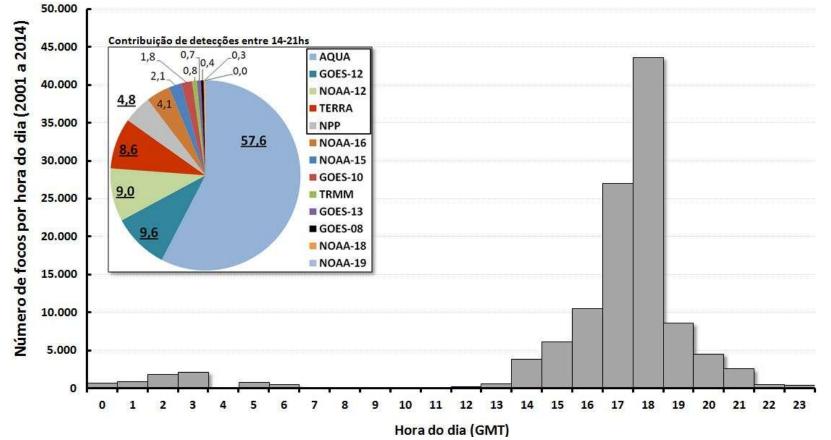
> *Sistema automático *Funcionamento 24h Courtesy: Alan Pimentel





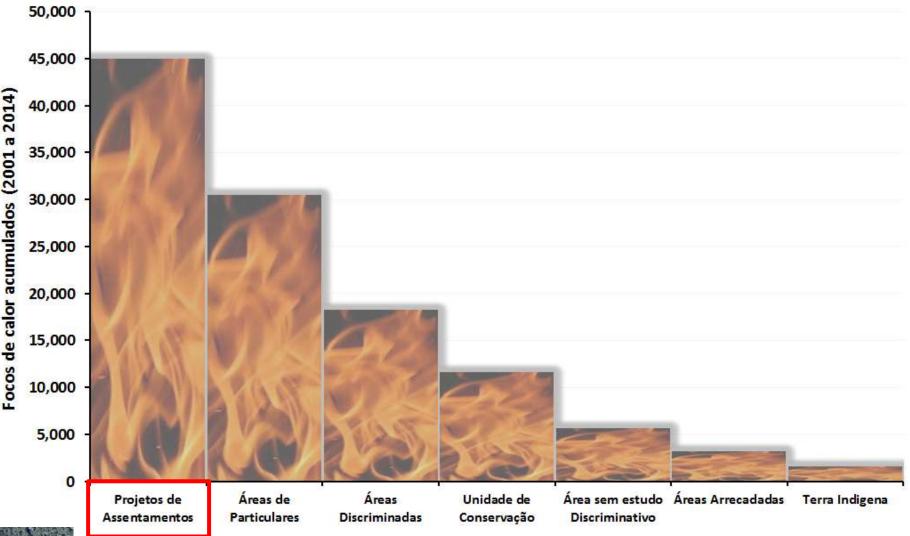


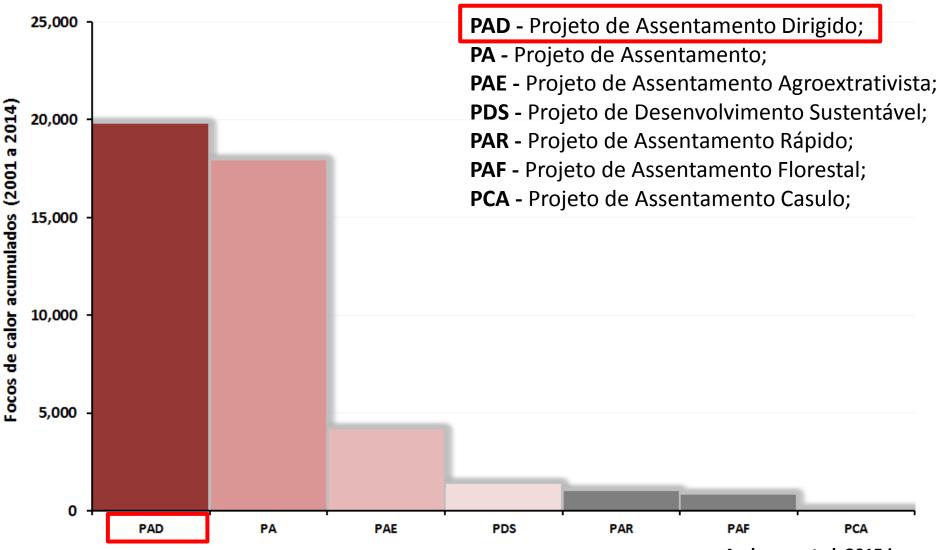
Pilot project: Acre State





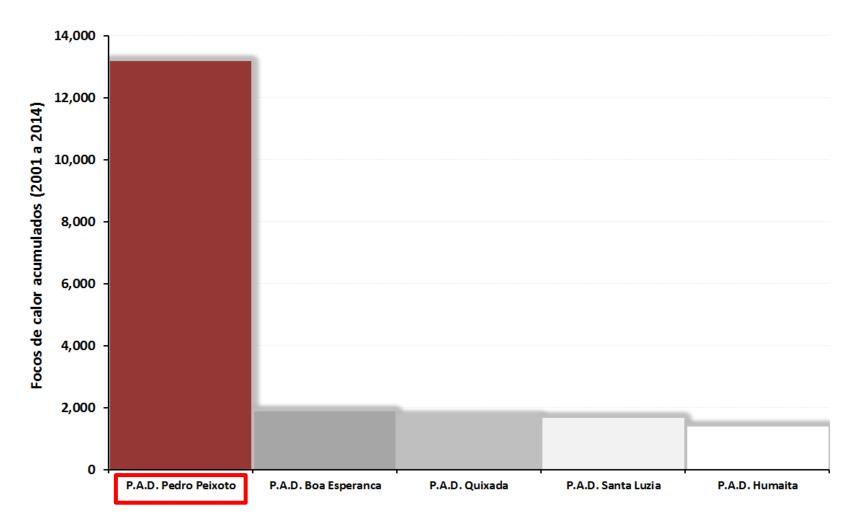
Pilot project: Acre State



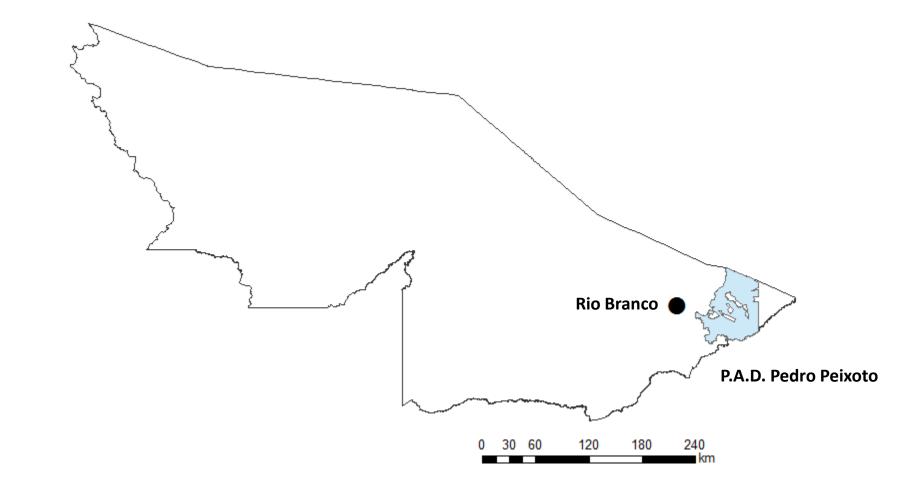


Anderson et al, 2015 in prep.

Pilot project: Acre State



3. Fire risk model





Pilot project: Acre State

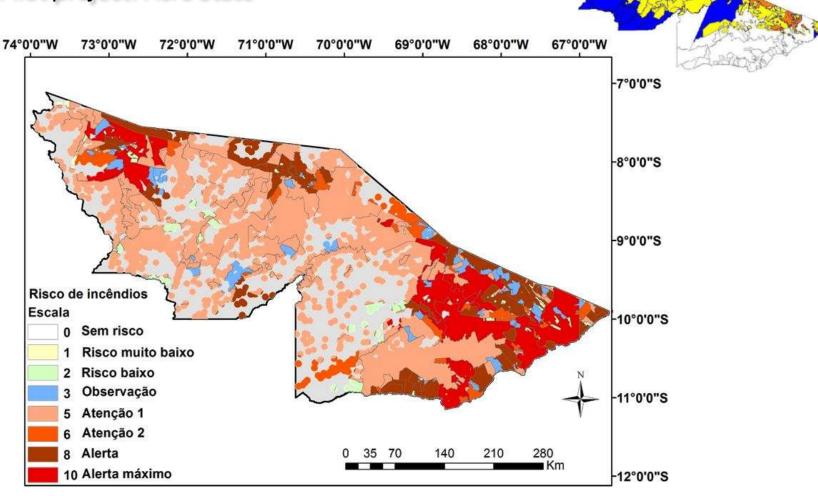
Analise de tendencia + analise do historico de uso (long-term e recente)

TABELA 3 Tabela de ordenamento de risco de incêndios baseando-se somente no histórico de ocorrência de fogo em cada unidade fundiária.

Risco de					
Incêndio	Escala	Observação			
		Tendência de aumento de focos de calor tanto entre 2000 e 2014 quanto entre 2010 e			
		2014, p<0.05			
		Ou			
		Tendência de aumento ou diminuição de focos de calor entre 2010 e 2014 não			
Alerta		significativa			
máximo	10	a 99.5%, mas com mais de 500 ocorrências de focos de calor entre 2010 e 2014.			
		Tendência de aumento de focos de calor tanto entre 2000 e 2014 quanto entre 2010 a			
		2014, p<0.1			
		Ou			
		Tendência de aumento ou diminuição de focos de calor entre 2010 a 2014 não			
		significativa a 99.5%, mas com 100 – 499 ocorrências de focos de calor entre 2010 e			
Alerta	8	2014.			
		Tendência de aumento de focos de calor entre 2010 e 2014, p<0.1			
		Ou			
Atenção 2	6	Entre 60 a 99 observações de focos de calor entre 2010 e 2014.			
		Tendência de aumento de focos de calor entre 2000 e 2014, p<0.1			
		Ou			
Atenção 1	5	Entre 41a 59 observações de focos de calor entre 2010-2014.			
Observação	3	Entre 11 e 40 focos entre 2010-2014.			
Risco baixo	2	até 10 focos entre 2010-2014			
Risco muito					
baixo	1	até 5 focos entre 2006-2014			
		Sem observação de focos de calor na série histórica (2000 a 2014)			
Sem risco	0	Ou a mais de 5 km de distância de qualquer foco de calor registrado entre 2000 e 2014.			

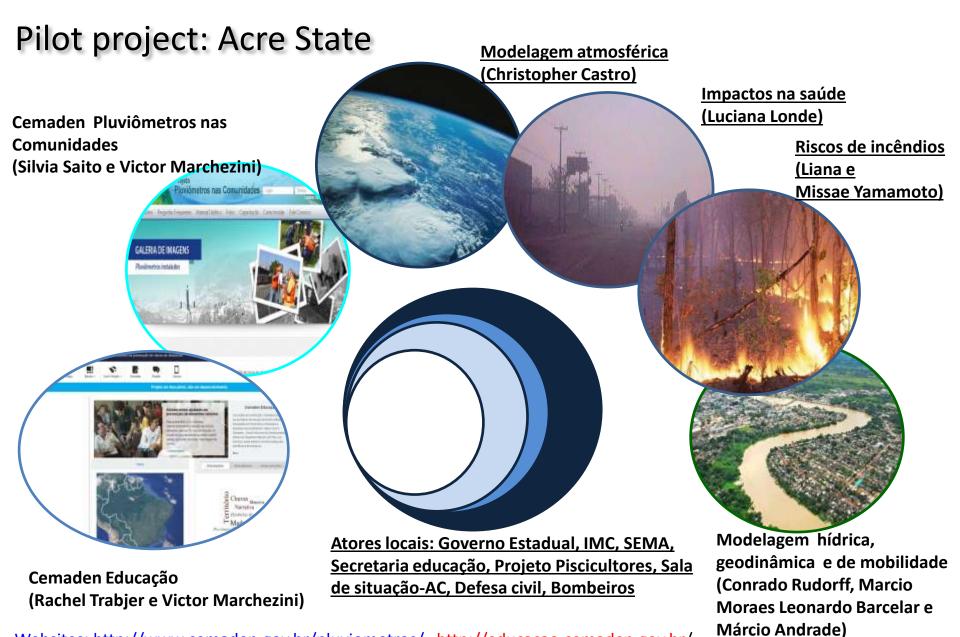


Pilot project: Acre State



Fire risk as a product of the observations of fire pixels: trend, historical use and number of observations Anderson et al, 2015 in prep.

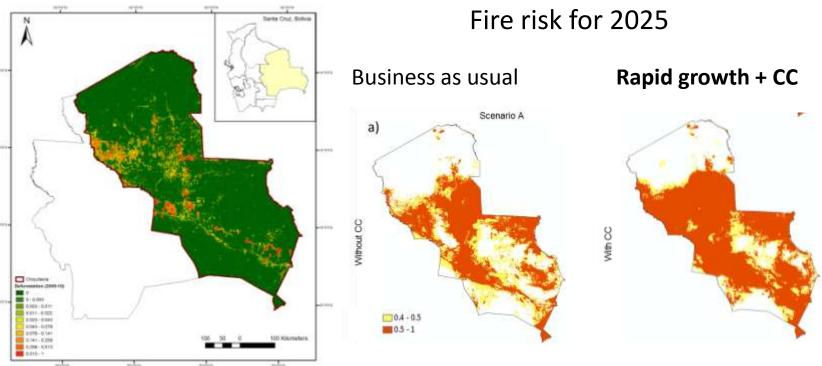
Integrative multidisciplinar project at Cemaden



Websites: http://www.cemaden.gov.br/pluviometros/ http://educacao.cemaden.gov.br/



Pilot project: Chiquitania region in the Department of Santa Cruz, Bolivia



important determinants of wildfire risk in this region are distance to roads, deforestation and density of human settlements. Severely dry conditions alone increased the area of high wildfire risk by 69%, affecting all categories of land use and land cover. the interactions between dry climatic conditions and rapid frontier expansion can increase the risk of wildfire even further.

Devisscher, Anderson et al, 2015 sub PLoS

Perspectives

What should be done for improving the understanding and the monitoring of fire and its associated impacts?

- 1. Burned area:
- Our product under development at TREES laboratory INPE: improved detection of forests affected by fires
- CPTEC in developing a burned area product for the cerrado region based on Landsat images, potentially for other areas in SA
- > Other initiatives: local, regional and global

2. Fire associated GHG emissions

- Increase the network of forests impacted by fires monitoring plots in SA for long term C stock assessment
- Improve the spatial representation of biomass decomposition decay
- Implement the emissions and C loss of forest areas affected by multiple fires

3. Fire risk

- To improve the understanding of local historical and economic drives for the use of fire
- To introduce local programmes for the understanding, adapting and mitigating the occurrence and impacts of fires
- Increase policy enforcement and monitoring systems



http://www.liana-anderson.org/



http://trees-research.weebly.com/



Fire on Starry Night Ronald Guerin