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LABORATÓRIO DE ANÁLISE E PROCESSAMENTO DE IMAGENS DE SATÉLITES



2. Remote sensing applications for dryland management in Brazil

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Managing Water Resources in Arid and semi-Arid Regions of Latin America and the Caribbean =____







THE COMPLEXITY OF DROUGHT QUANTIFICATION AND ANALYSIS

• Droughts are difficult to pinpoint in time and space given <u>different</u> <u>economic sectors and natural systems affected.</u>

• We identify a drought by its effects or impacts on different types of systems (agriculture, water resources, ecology, forestry, economy, etc.), but there is not a physical variable we can measure to quantify droughts.

• Long-term drought objective metrics (streamflows, soil moisture, lake levels, etc.) are commonly not available. Moreover, using only objective metrics other <u>relevant variables</u> to determine drought severity (e.g. <u>the atmospheric water demand</u>) are not taken into account.

• We use the so-called <u>"DROUGHT INDICES"</u> for drought quantification and analysis.



EXISTING DROUGHT INDICES

Precipitation-based drought indices, including the SPI, rely on two assumptions:

- 1) The variability of precipitation is much higher than that of other variables, such as the atmospheric water demand
- 2) The other variables are stationary (i.e., they have no temporal trend).
- In this scenario, the importance of these other variables is negligible, and droughts are controlled by the temporal variability in precipitation.

Is this scenario plausible nowadays?



The Challenge and Our Goals

The challenge: How can early warning of physical drought conditions be provided in regions with low data availability and low capacity?

The goal: To develop a flexible framework for providing historic records, real-time monitoring and forecasts of relevant hydrological variables and drought indices

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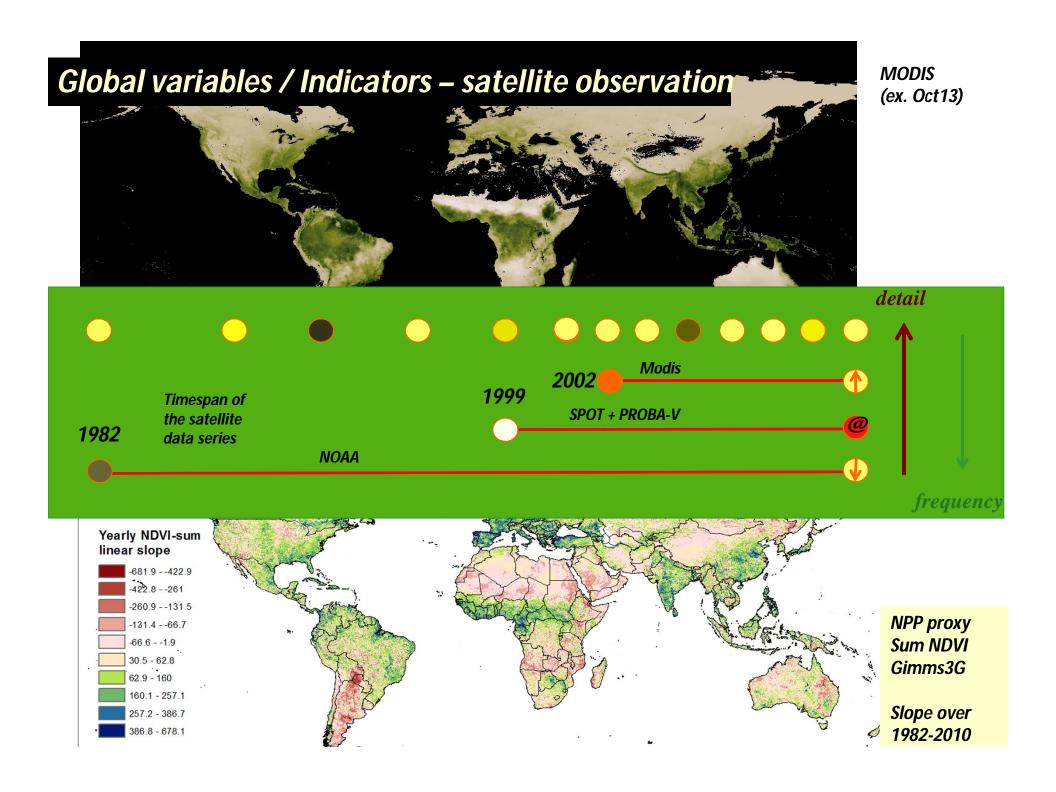
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Background

- Vegetation over a ground, forms naturally or by cultivation attained a great importance in the context of changing climate scenario over the globe.
- Vegetation is also controlled by the climate which in turn leads to climate vegetation feedback mechanism.



NDVI

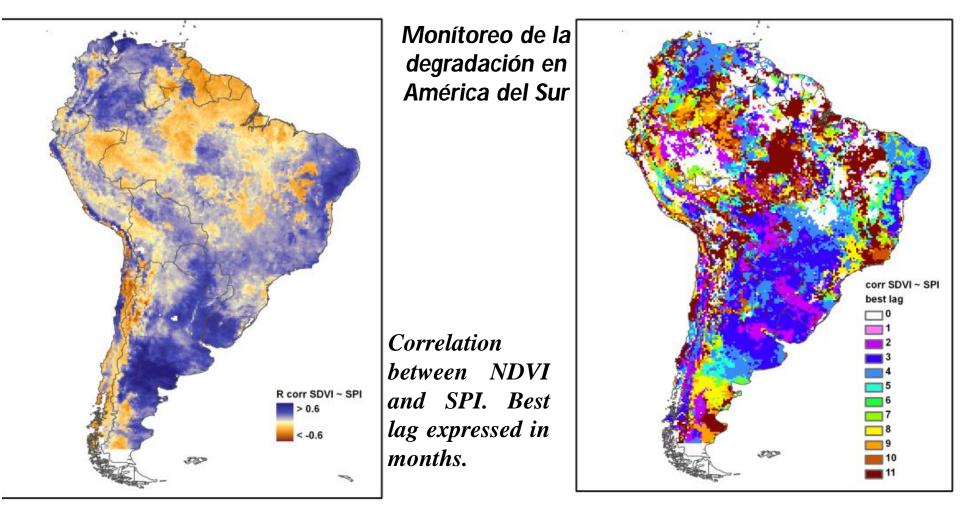
• NDVI is calculated from the visible and nearinfrared light reflected by vegetation.

 $NDVI = \frac{NIR - RED}{NIR + RED}$

- Healthy vegetation (left) absorbs most of the visible light that hits it, and reflects a large portion of the nearinfrared light.
- Unhealthy or sparse vegetation (right) reflects more visible light and less near-infrared light.
- NDVI valves from -1 to +1.

NDVI < 0 ----0 < NDVI < 0.2 ----0.2 < NDVI < 0.4 ----0.4 < NDVI < 0.8 ----NDVI > 0.8 ----

- - Water Bodies Less Vegetation Medium Vegetation High Vegetation Rain Forest



SPOT-Vegetation 10 compuesto diario NDVI de datos (desde abril 1998 hasta marzo 2012) de 1 km (http://www.vgt.vito.be/)

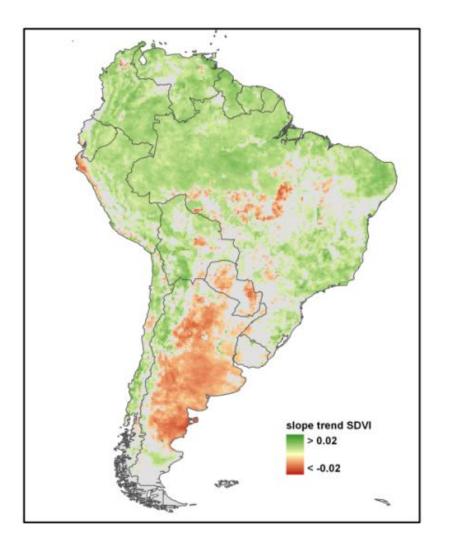
10-precipitación diaria en 0,25 °, disponible en (ECMWF)

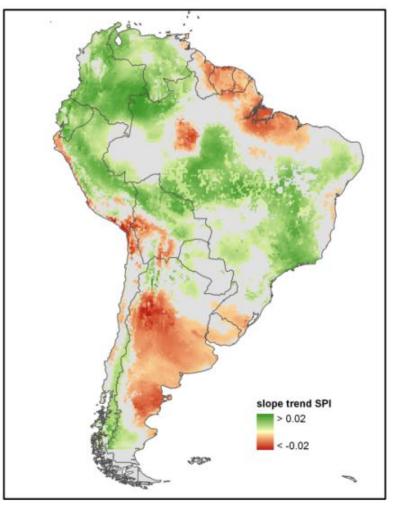
Standardized Difference Vegetation Index (SDVI)

Standardized Precipitation Index (SPI)

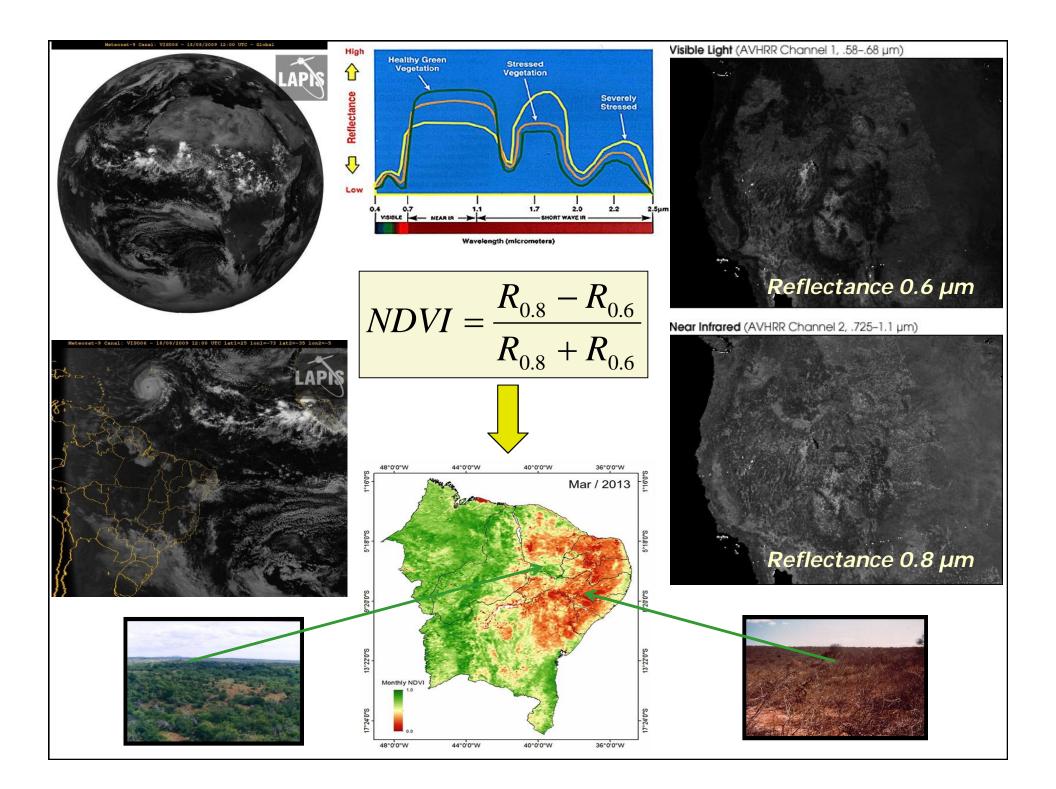
Fuente: Barbosa et al. (2012)

Slope of the trend analysis of SDVI (left) and SPI (right). Non significant trends are masked in grey.

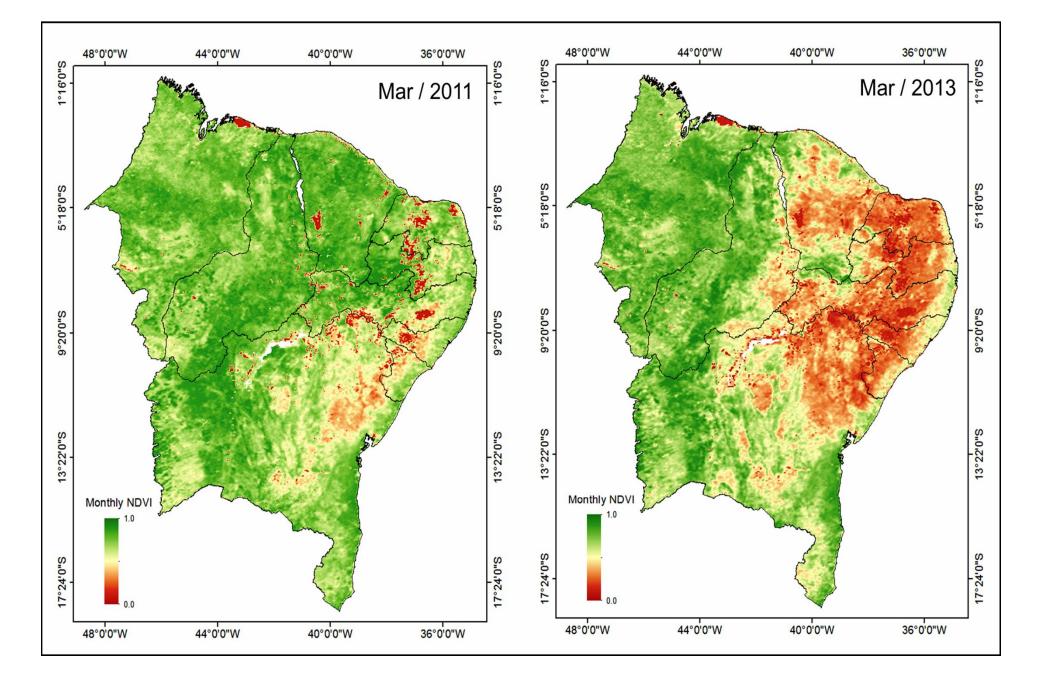




Fuente: Barbosa et al. (2012)



LAPIS e INSA: (MSG NDVI)



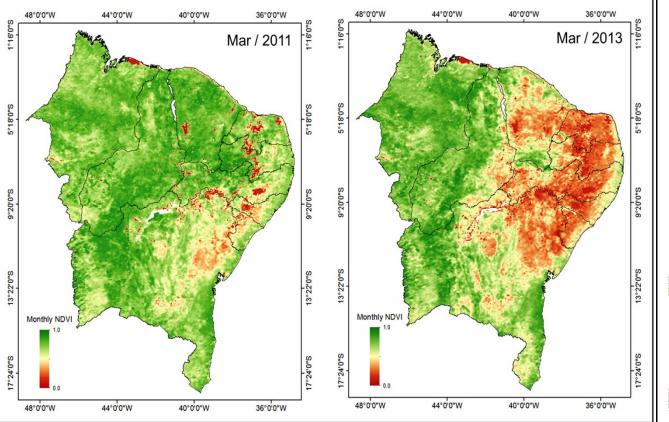
LAPIS and INSA: MSG NDVI

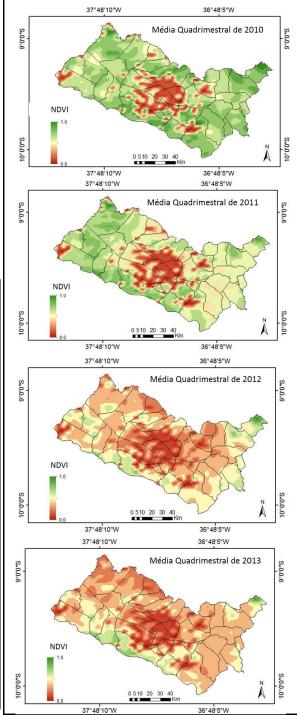


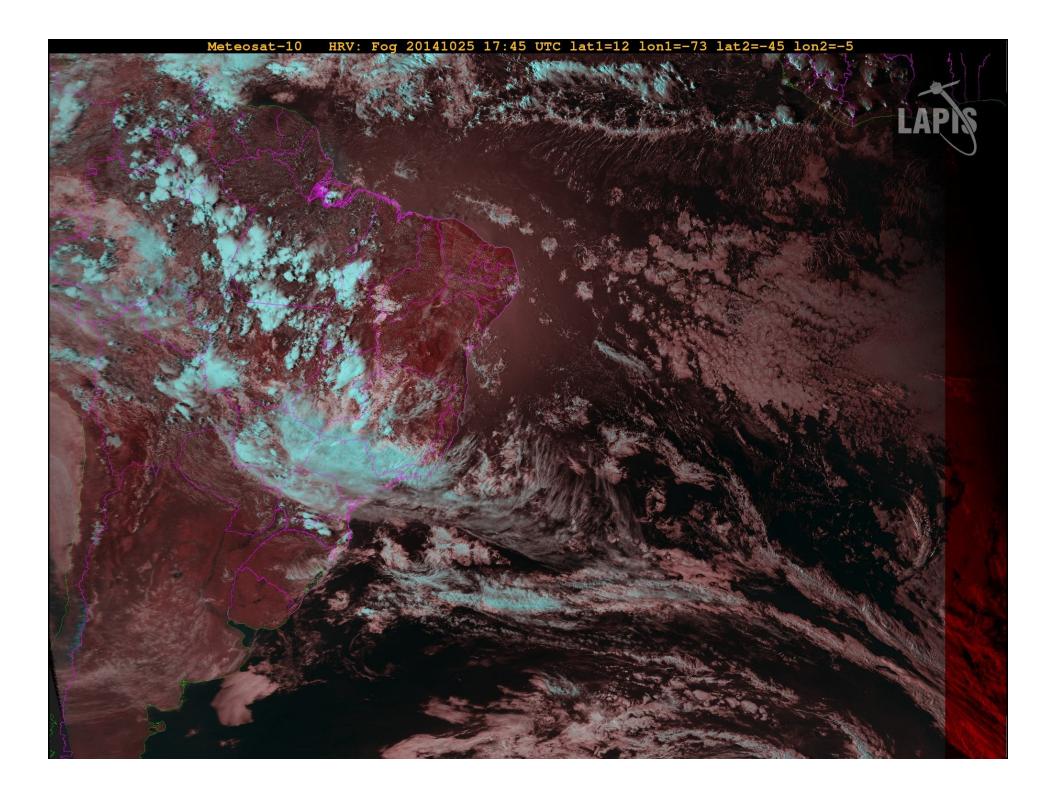
Alagoas

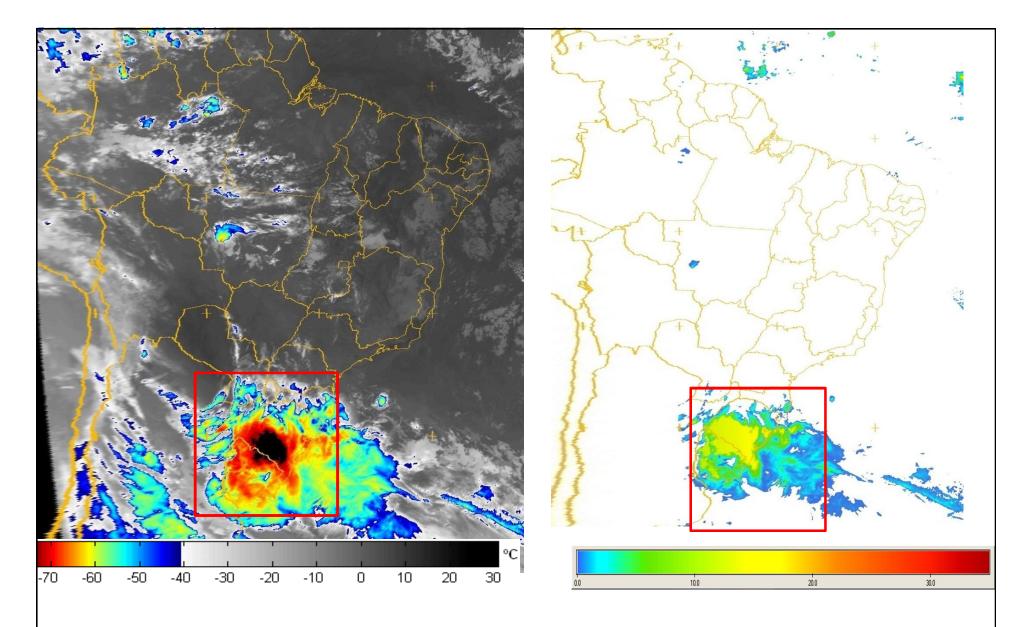


Nordeste



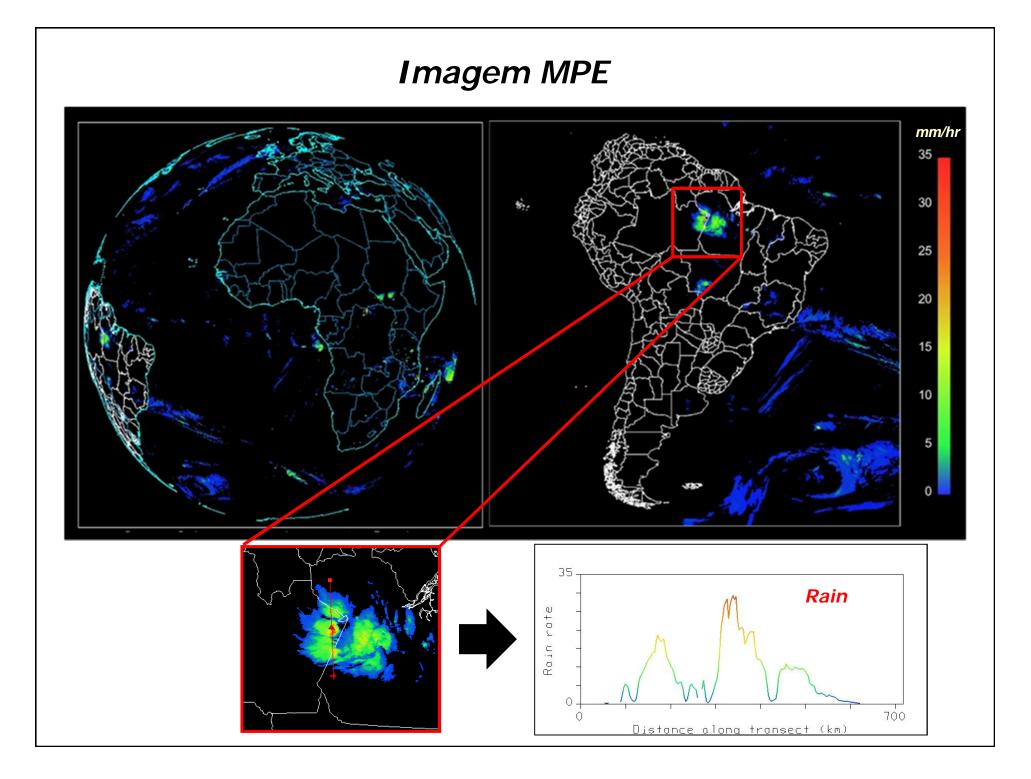




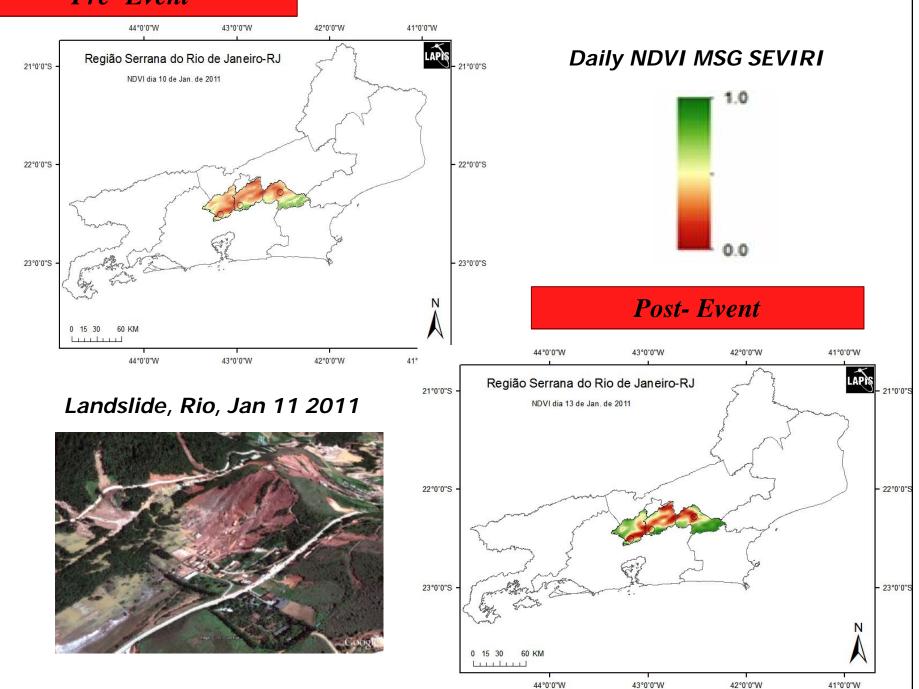


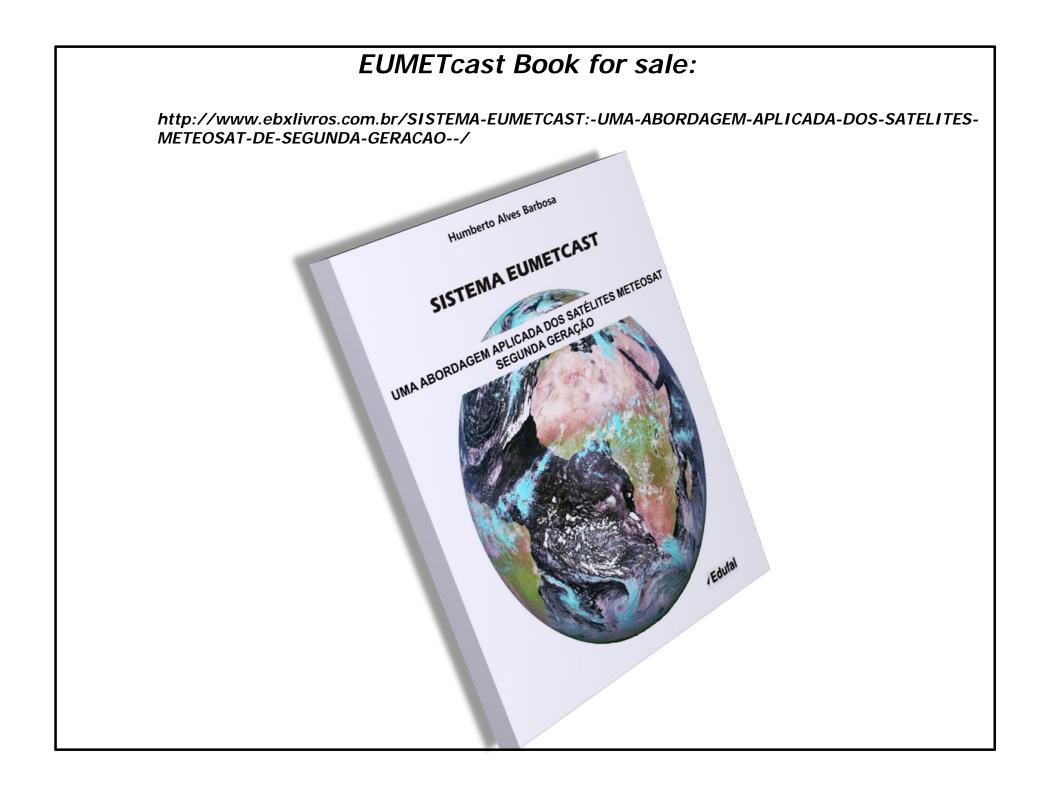
: MSG IR10.8

Product: MSG MPE Rainfall



Pre- Event





Questions

Q1) What types of drought/flood tools or research has your institution funded, and how has that helped achieve your priorities.

Q2) What do you think are the great gaps in your priorities that drought/flood "science and tools" could help close?

Q3) What strategies are there to enhance the drought/flood science - governance linkages?



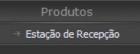
The End

LABORATÓRIO DE ANÁLISE E PROCESSAMENTO DE IMAGENS DE SATÉLITES

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Links



•Thank you for listening!

O Laboratório de Análise e Processamento de Imagens de Satélites (LAPIS) da Universidade Federal de Alagoas (UFAL) realiza atividades de pesquisa, assistência tecnológica e treinamento de recursos humanos para a recepção, processamento, interpretação e integração de imagens dos satélites da série METEOSAT. Para atender a essa demanda, em 2007 a UFAL instalou e operacionalizou a terceira estação de recepção de imagens do satélite METEOSAT Segunda Geração (MSG) do Brasil. Como atividades de pesquisa e transferência de conhecimento, a equipe do LAPIS elabora aplicativos para tratamento de imagens, disponibiliza produtos meteorológicos e ambientais derivados do MSG para setores operacionais e oferece treinamento na área. Desenvolvidas inteiramente com ferramentas open-source e freeware.

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→ 2006

··> 2007

> 2008

2009