

# DETECTING CHANGE FROM DENSE TIME-SERIES

## **BFAST**

Breaks For Additive Season and Trend



07-11 May 2018 | Phnom Penh







## Introduction

Why are we all here?

Improve monitoring and quantification of forest loss and forest gain

Why are we interested in time series/time series analysis?











#### What is BFAST?



- Bfast is an open source tool
- Countries can use it themselves, adapting it to their needs









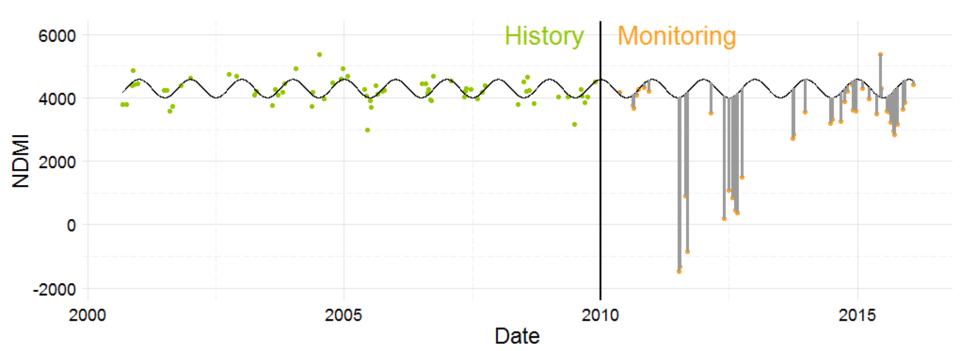
For this, the MOSUM test is applied

Testing the stability of the model = to check if the "new data" still "fits" the model

If the data still fits => no change is detected else => a break is marked

## **BFAST Monitor**

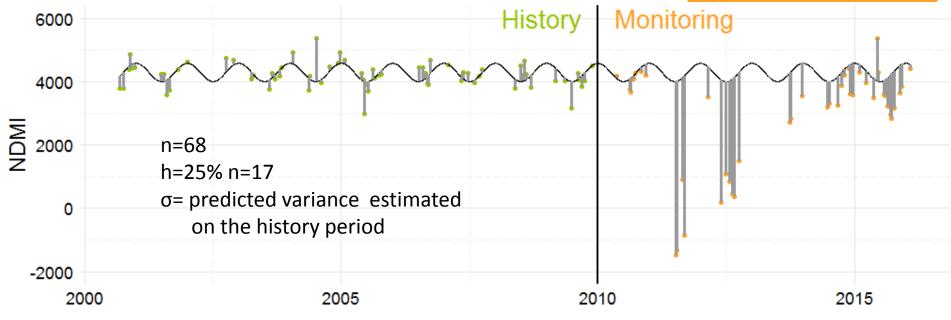
The method consists in fitting a model to the data by Ordinary Least Squares (OLS) fitting, on a period defined as stable history, and testing for stability of the same model, during a period defined as monitoring period.



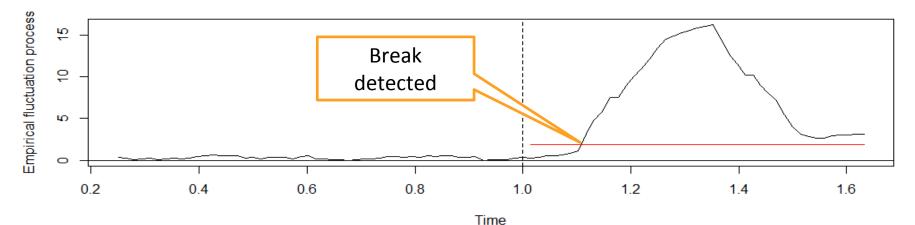
## **Applying MOSUM**

$$MO_t = \frac{1}{\hat{\sigma}\sqrt{n}} \sum_{s=t-h+1}^{t} (y_s - \hat{y}_s)$$

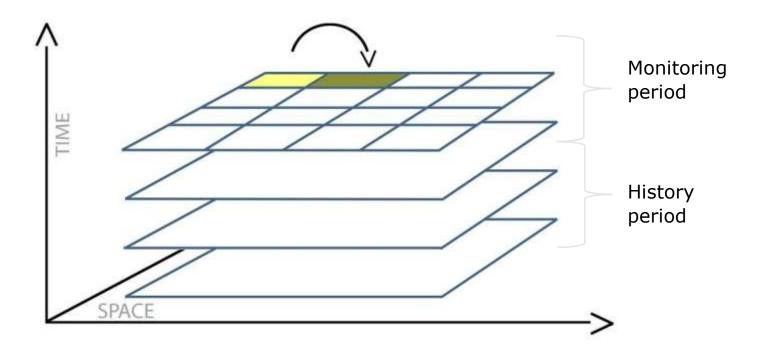
MOSUM detects a break when the MO passes the 95% significance boundary



#### Monitoring with OLS-based MOSUM test

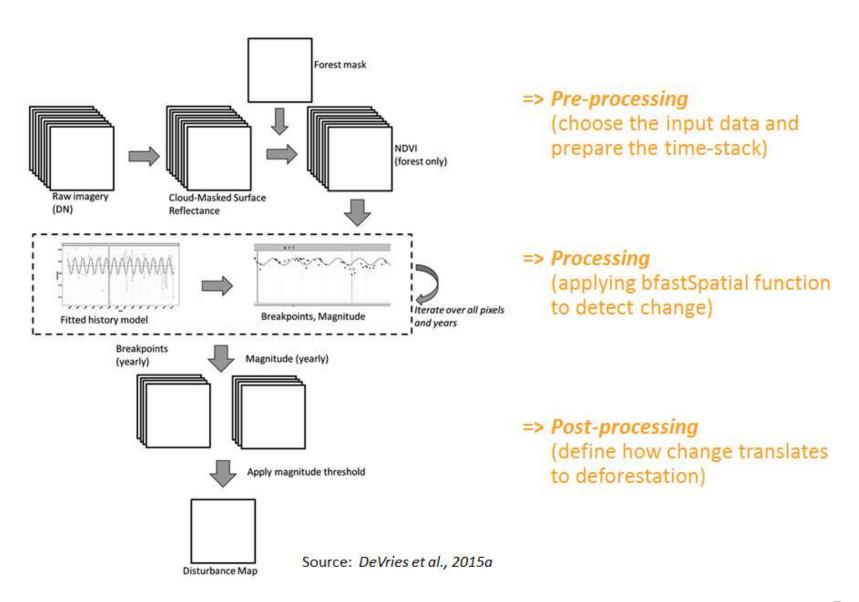


## **BFAST Spatial**

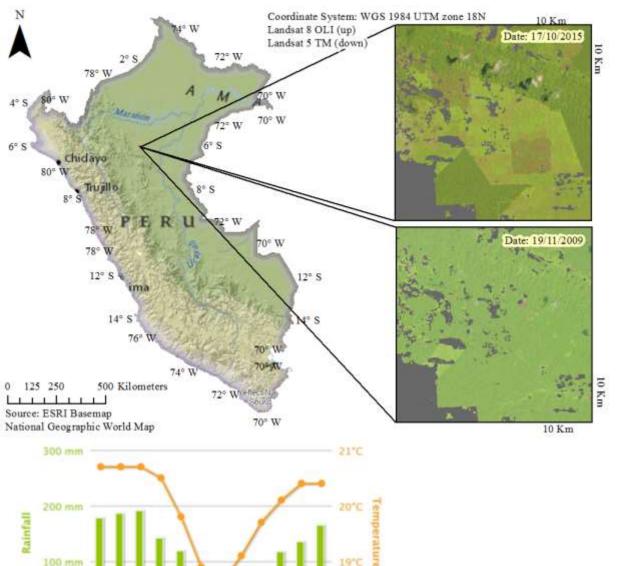


- all individual time series (one pixel) WILL have the same monitoring period
- all individual time series (one pixel) CAN have the same history period
- each individual time series (one pixel) WILL have the same regression model (ex:harmonic order 1), but different parameters
- Change detection is unique for each pixel

## Analysis overview: from Landsat scenes to deforestation detection



## Case study: Peru



#### **Forest Characteristics**

- Evergreen forest with seasonality: "selva alta"
- Loreto is the largest region in Peru and has had more hectares deforested than any other (Peruana, 2015 "Revealing the hidden")
- Deforestation in large "blocks resulting from agricultural investments (mainly palm oil)



Source: Barranquita Resiste<sup>2</sup>

<sup>1</sup>http://sdwebx.worldbank.org/climateportal/ <sup>2</sup>https://cordilleraescalera.wordpress.com/bbarranquita-resistsb-2/

Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

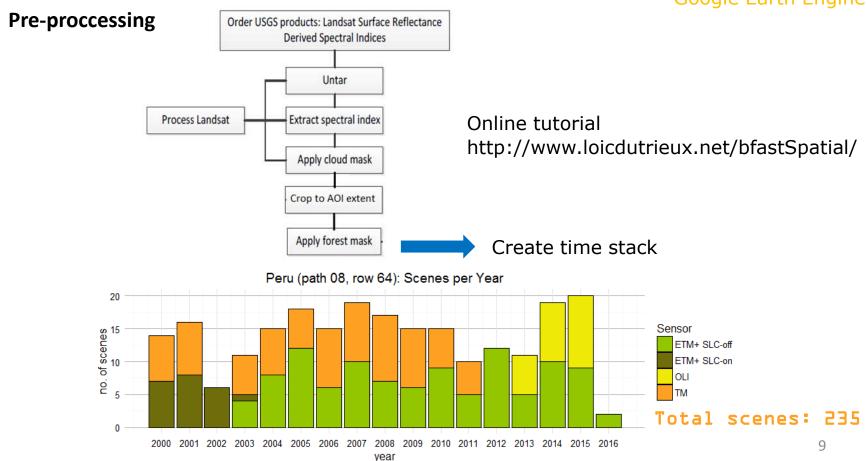
## **Data acquisition and Pre-processing**

#### **Products acquired:**

- NDVI and NDMI derived from Surface Reflectance products
- Cloud Mask
- Forest Mask (for 2010)

processed and provided by the USGS

Google Earth Engine



## Discuss and understand the parameters of the bfastSpatial function

```
bfmSpatial(x, dates = NULL, pptype = "irregular", start, monend = NULL,
  formula = response ~ trend + harmon, order = 3, lag = NULL,
  slag = NULL, history = c("ROC", "BP", "all"), type = "OLS-MOSUM",
  h = 0.25, end = 10, level = 0.05, mc.cores = 1,
  returnLayers = c("breakpoint", "magnitude", "error"), sensor = NULL, ...)
```

Guide: https://github.com/rosca002/Testing\_BFAST\_settings/blob/master/Bfast\_Spatial\_Guide.md

#### 1. Input data: What vegetation index to use?

Convenient: NDMI or NDVI

Recommended: NDMI









#### 2. History Period

- (i) How long should the history period be?
  - depends on the regression model
  - frequently clouded areas: min 55-60 scenes in the history period
  - enough observations per pixel for the algorithm to fit a model (a min of 20 observations per pixel, and a mean of 40-50 observations per pixel)
  - the more cloudy the scenes are, the bigger the number of scenes needed.
- (ii) How to have a disturbance free history period?
  - options: all, ROC
  - a moment that delineates a stable period in the history period can be provided by expert knowledge
  - or can be calculated automatically using the reverse-order-cumulative sum (ROC or CUSUM) of residuals
  - extremely low number of scenes available (due to cloud coverage, e.g. Gabon) it is recommended to use all scenes available in the history period, with the condition to visually assess the study area for disturbances in this period







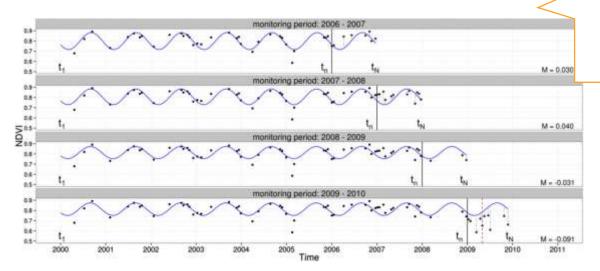


#### 3. Monitoring period

#### (i) Full monitoring period approach

```
bfmSpatial(ndmiStack, start = c(2010, 1), formula = response~harmon,
order = 1, history = c(2000, 1), filename = out))
```





Limits the monitoring period to one year

- the history period is enlarged with every iteration
- this approach can make a big difference in the cases with very few observations available

### 4. Regression model

#### (i) the phenology of the forest

- choose the harmonic order of the model to follow as closely as possible the seasonal patterns
- decide if trend is, or not, to be included in the model

#### (ii) the number and frequency of the cloud-free available imagery.

- the more complex the regression, more observations are needed in the history period
- even though present, complex seasonal patterns might not be detectable with Landsat data alone, if the AOI is frequently cloud covered



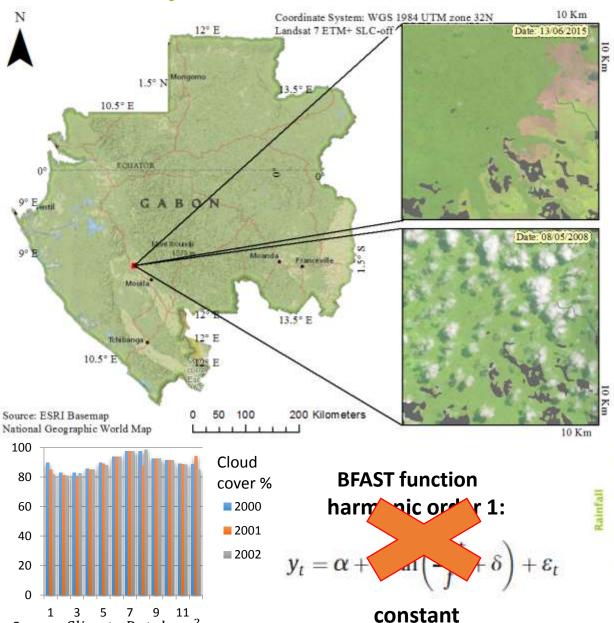




## **Case Study: Gabon**

Source: Climate Database<sup>2</sup>

<sup>2</sup>http://droppr.org/data/map/cru climateportal/



#### **Forest Characteristics**

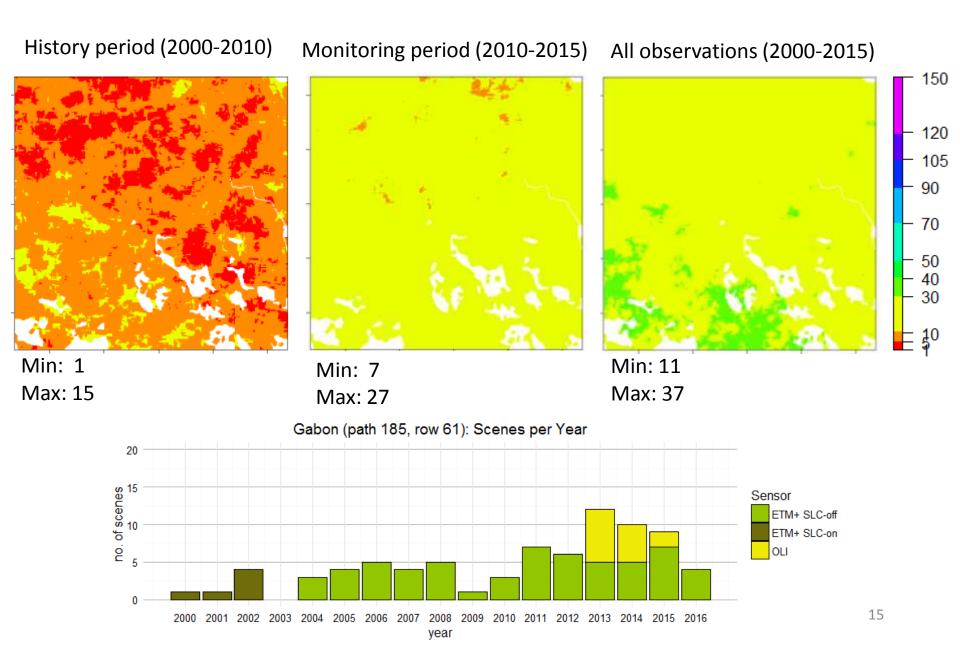
- Evergreen ecuatorial forest with no seasonality
- One of the most cloudy area on Earth



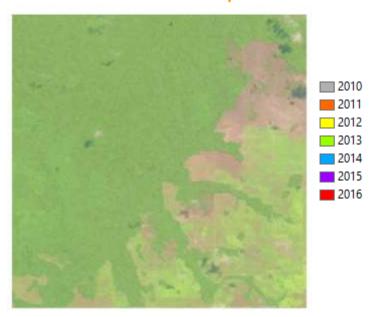
Source: Climate change Knowledge Portal<sup>1</sup>

1http://sdwebx.worldbank.org/climateportal/

## No. Observations per pixel



BFAST Seq













## Presentation adapted from Sabina Rosca

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www.un-redd.org www.unredd.net







