

→ EO OPEN SCIENCE 2017

#EO #OpenScience

VISUALISATION
TOOLBOXES
BLOGS
COLLABORATION
E-LEARNING
SCIENCE
OPEN-DATA
VIRTUAL-LAB
CITIZENS
CLOUD
MOOC

dtwSat

An R Package for Land Cover Classification Using Satellite Image Time Series

Victor Maus

International Institute for Applied Systems Analysis, IIASA
Laxenburg, Austria



Open-source software for satellite image time series analysis

Available from the **Comprehensive R Archive Network (CRAN)**
<https://cran.r-project.org/web/packages/dtwSat/index.html>

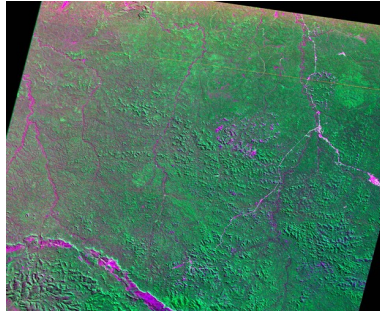


```
> install.packages("dtwSat")
```

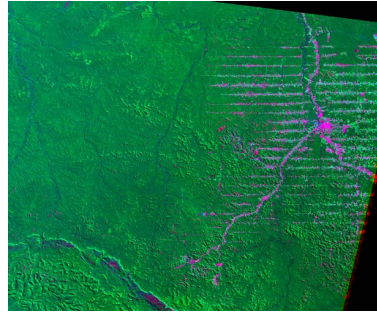
Development version on GitHub
<https://github.com/vwmaus/dtwSat/>



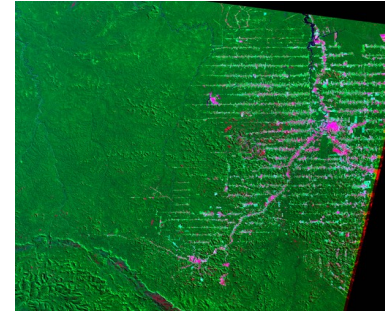
1975



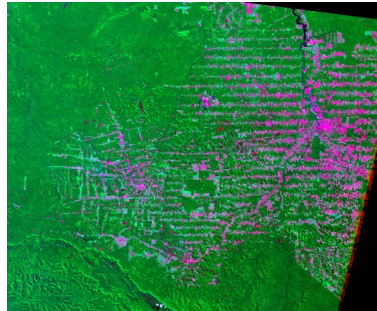
1986



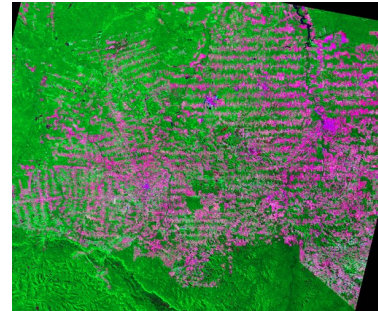
1992



2001

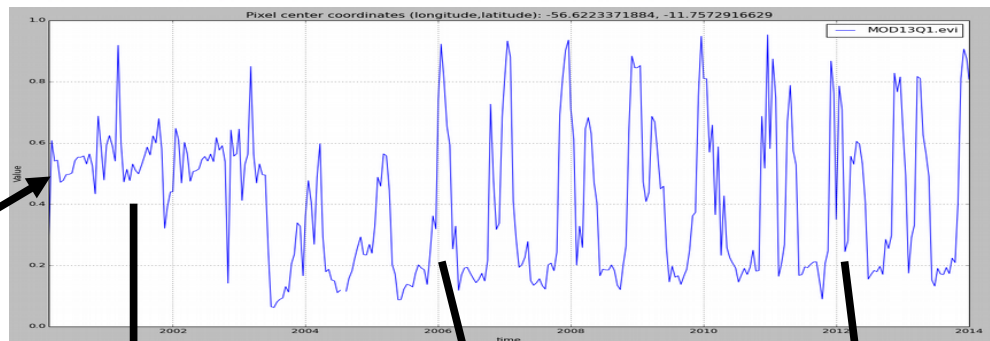
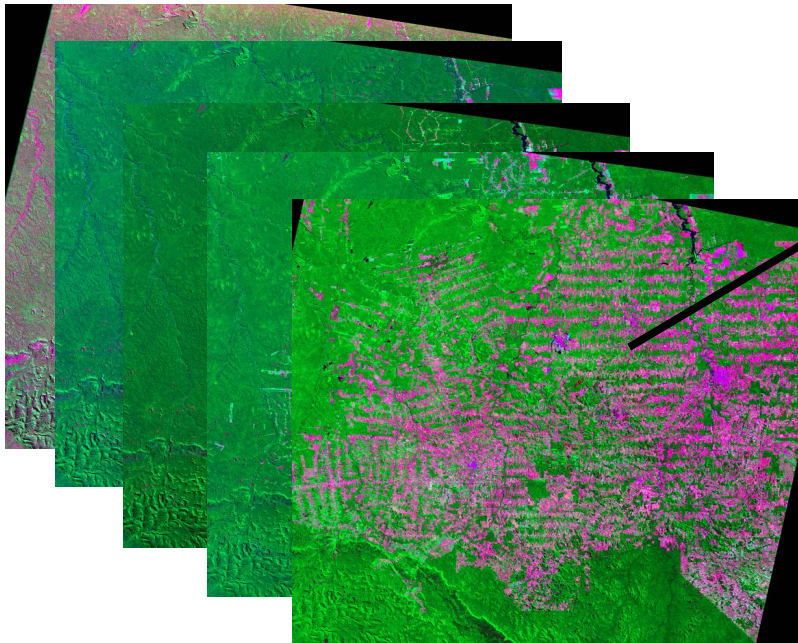


2013



Rondônia, Brazil

Images source: <http://earthshots.usgs.gov/earthshots/node/39#ad-image-4>

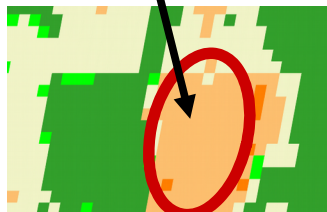


Forest



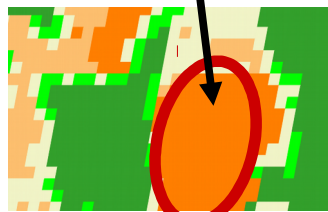
2001

Single cropping

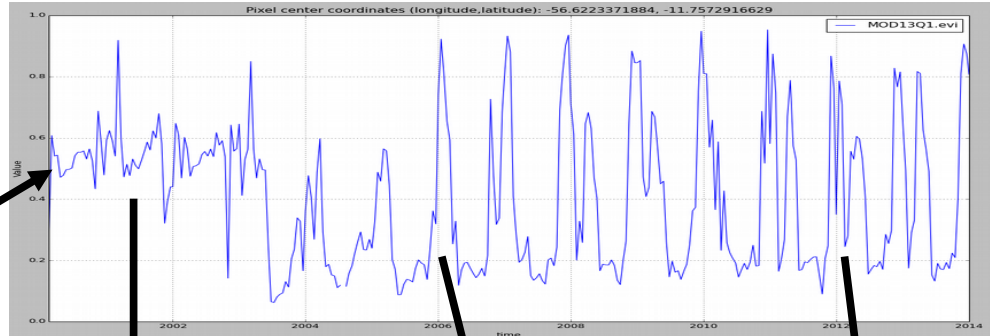
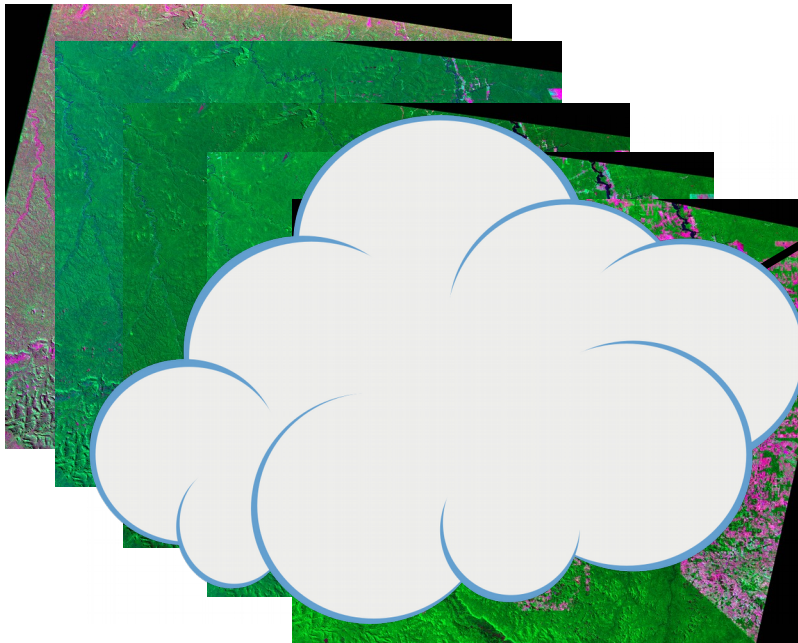


2006

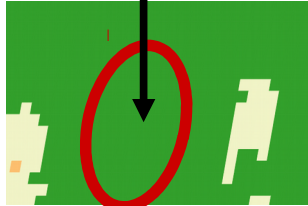
Double cropping



2013

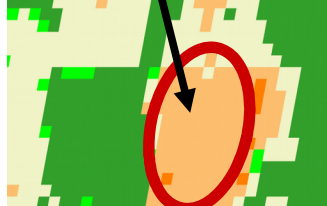


Forest



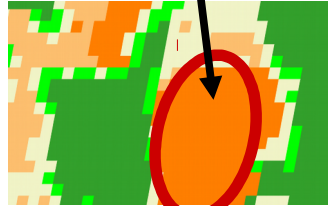
2001

Single cropping



2006

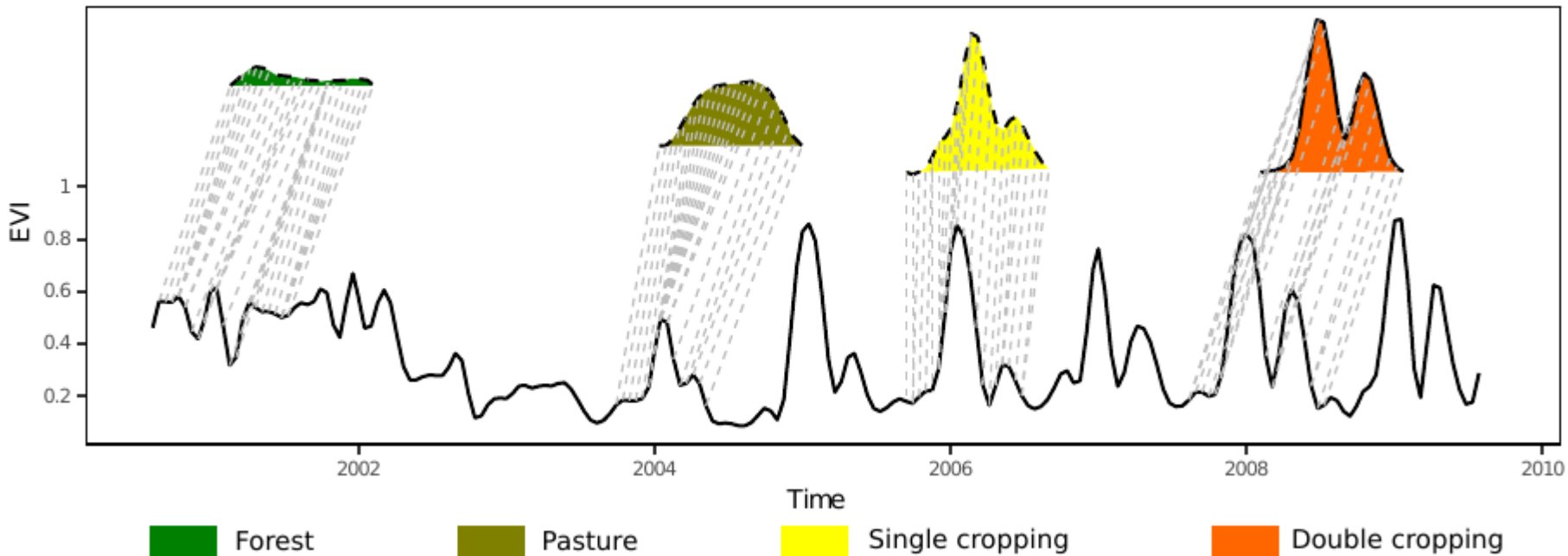
Double cropping



2013

Usually satellite time series are irregularly sampled, noisy, and out-of-phase.

A good match needs shape similarity and temporal coherence



Time-Weight Dynamic Time Warping (TWDTW) method Maus et al., (2016)



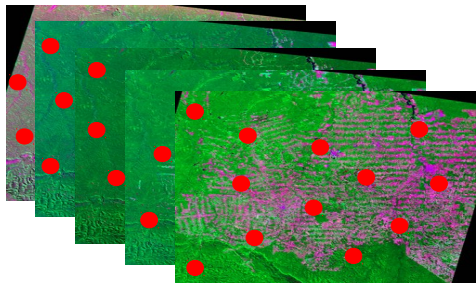
Sandstone sculpture in the old Laxenburg train station, Austria.
St. Mark's Lion is the symbol of the city of Venice.



Unknown

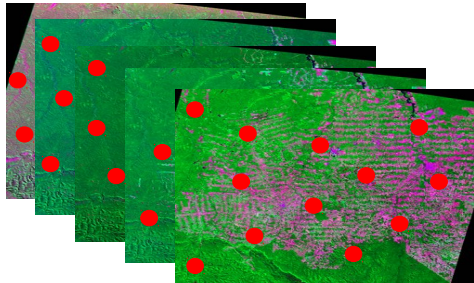
INPUT

1. Satellite Time Series
2. Ground truth samples



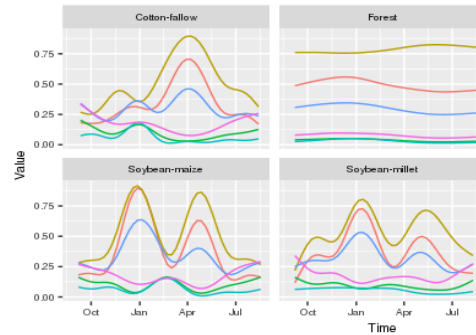
INPUT

1. Satellite Time Series
2. Ground truth samples



GAM
→

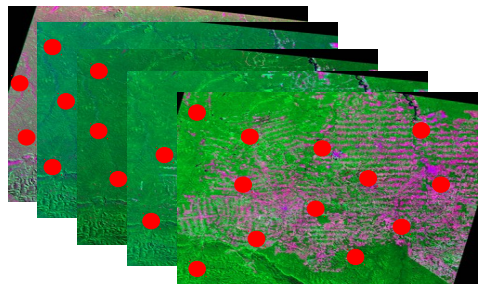
Temporal patterns



How does dtwSat work?

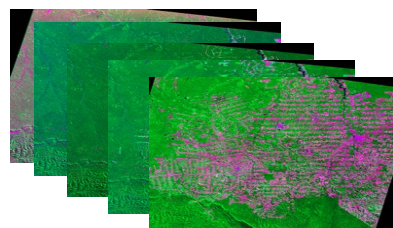
INPUT

1. Satellite Time Series
2. Ground truth samples

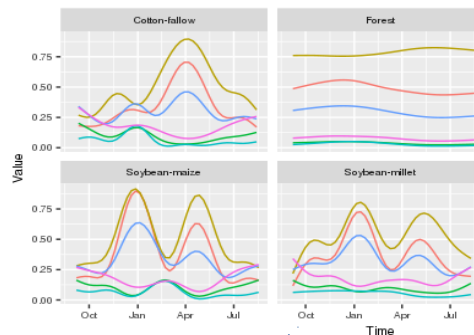


GAM

Raster stack

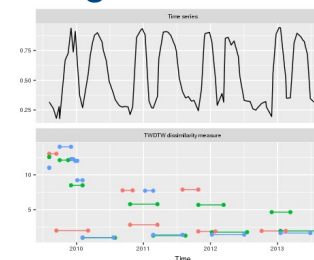


Temporal patterns



TWDTW

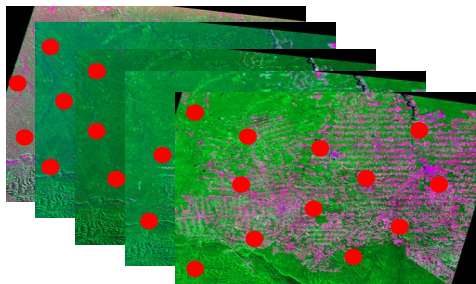
TWDTW alignments



How does dtwSat work?

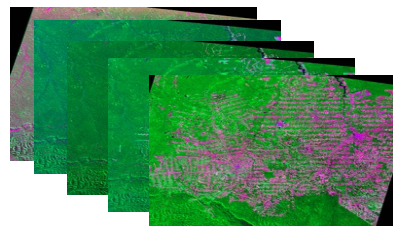
INPUT

1. Satellite Time Series
2. Ground truth samples

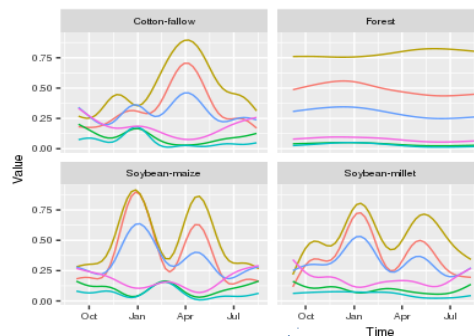


GAM

Raster stack



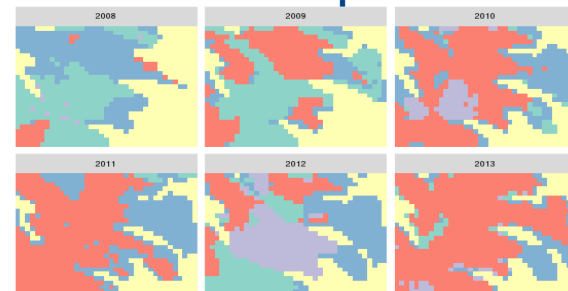
Temporal patterns



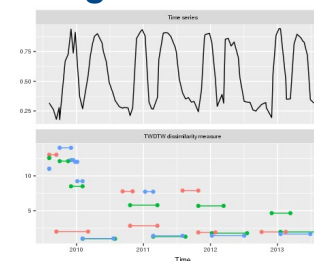
TWDTW

OUTPUT

Time series of maps



TWDTW alignments





11°59'08"S
55°59'51"W

Study Area

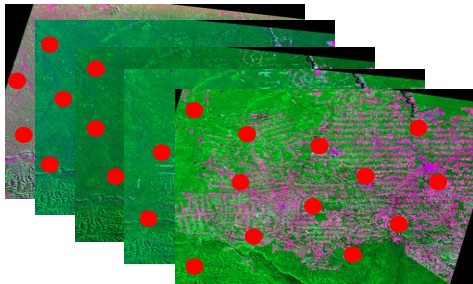


12°02'37"S
55°54'33"W



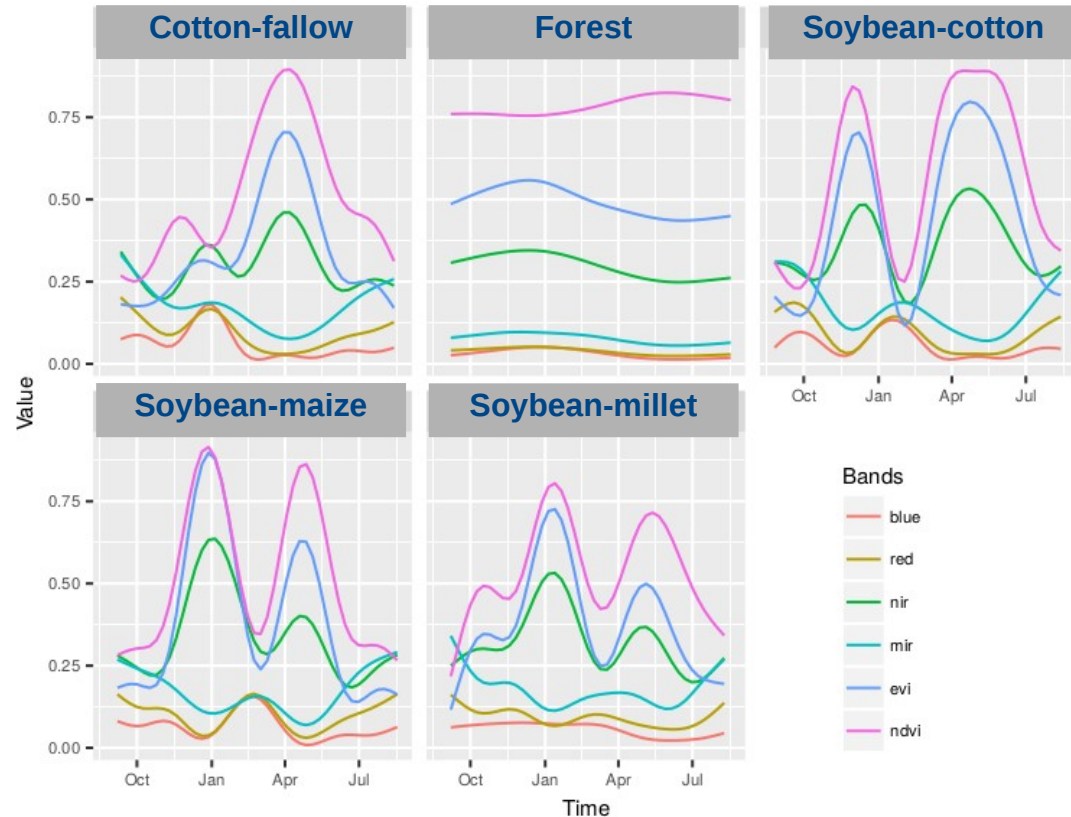
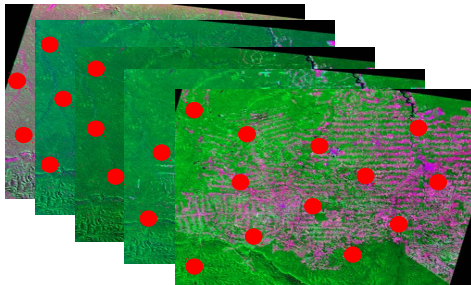
INPUT

1. MODIS MOD13Q1 Time Series
Bands: blue, red, nir, mir, evi, ndvi
2. Ground truth samples

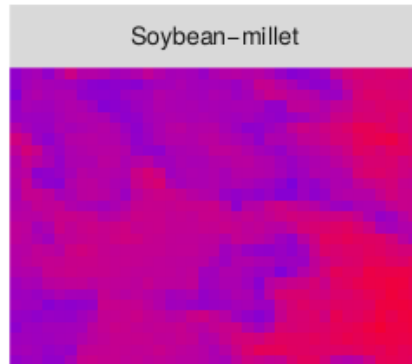
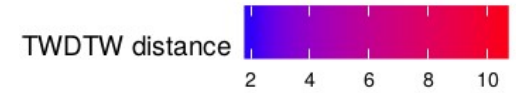
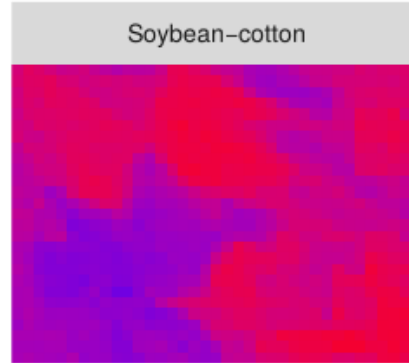
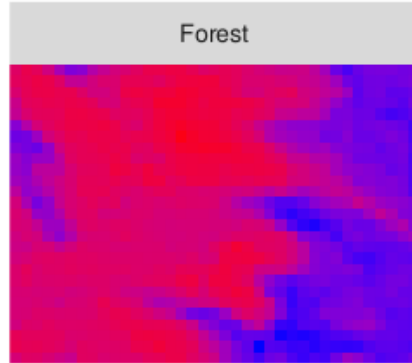
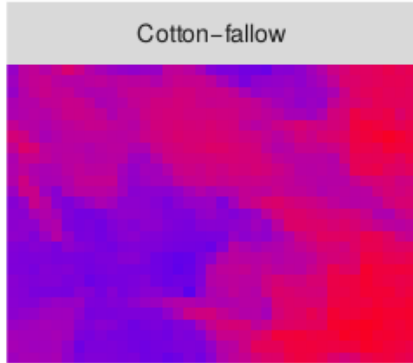


INPUT

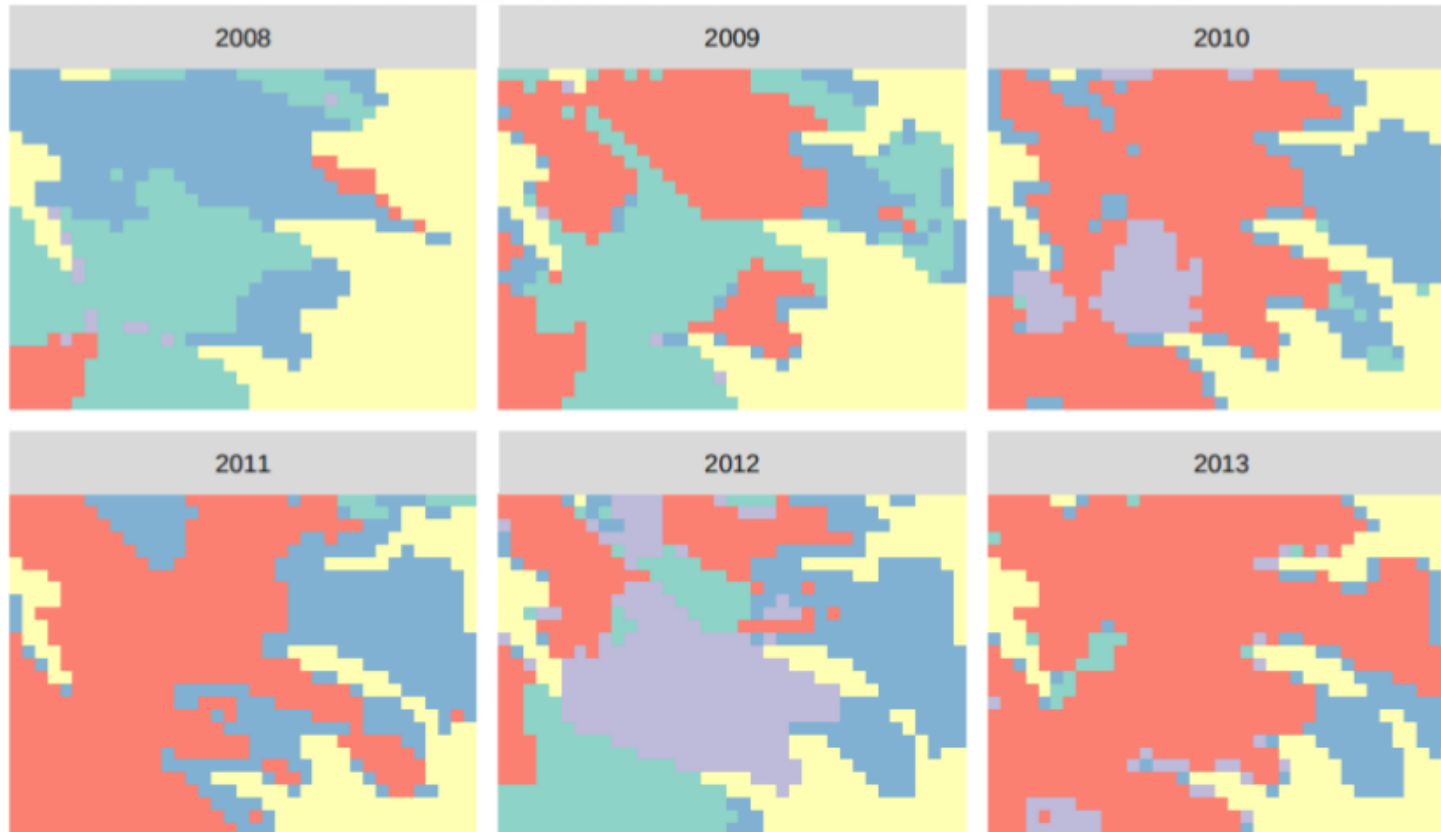
1. MODIS MOD13Q1 Time Series
Bands: blue, red, nir, mir, evi, ndvi
2. Ground truth samples



2008



Overlapping the layers we find the most similar class for each pixel.



MODIS

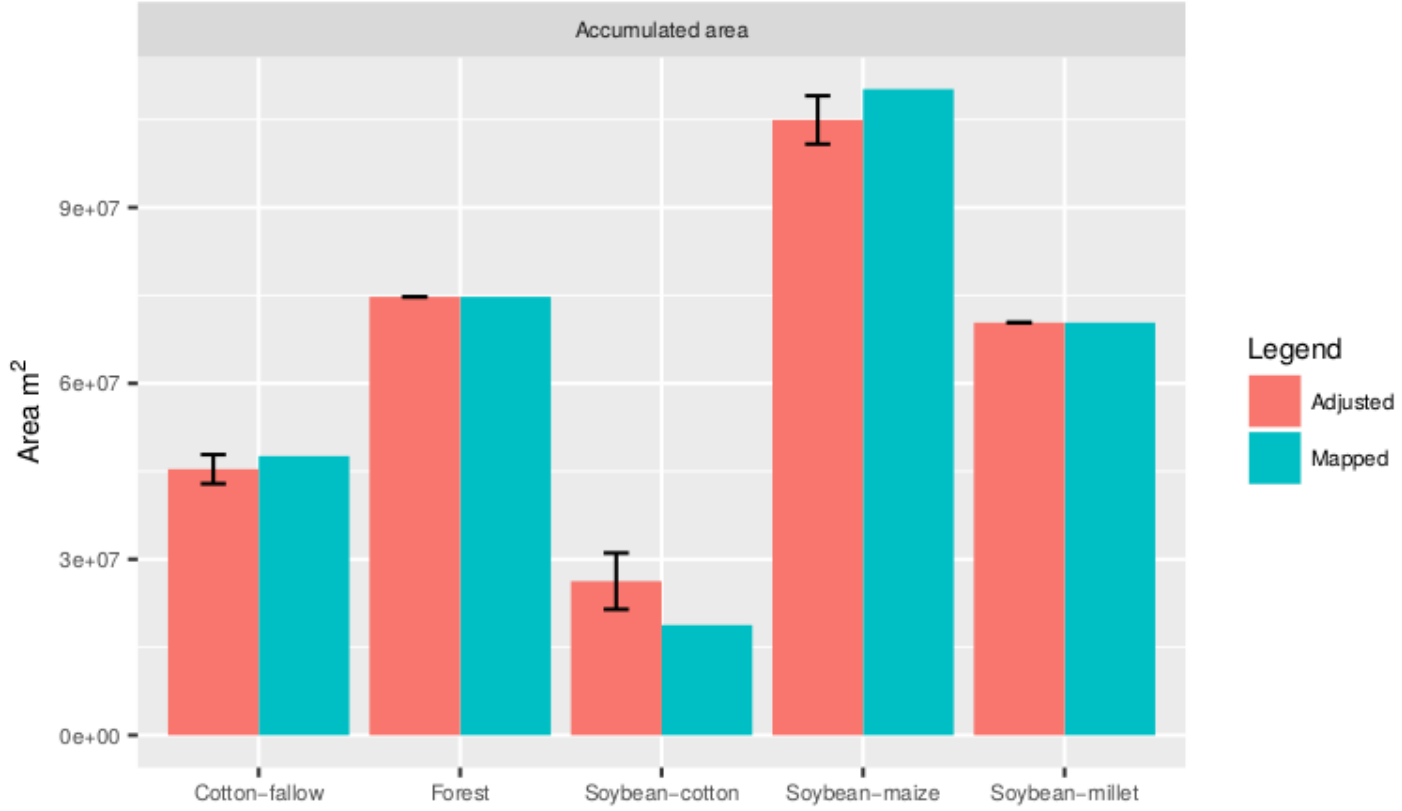
- Cotton-fallow
- Forest
- Soybean-cotton
- Soybean-maize
- Soybean-millet
- unclassified

Author | ESRIN | 18/10/2016 | Slide 17



Map class	Reference class					Total	User's*	Producers's*	Overall*
	Cotton-fallow	Forest	Soybean-cotton	Soybean-maize	Soybean-millet				
Cotton-fallow	0.14	0.00	0.00	0.00	0.00	0.14	0.95±0.05	1.00±0.00	0.98±0.01
Forest	0.00	0.23	0.00	0.00	0.00	0.23	1.00±0.00	1.00±0.00	
Soybean-cotton	0.01	0.00	0.06	0.02	0.00	0.08	1.00±0.00	0.72±0.13	
Soybean-maize	0.00	0.00	0.00	0.33	0.00	0.33	0.95±0.04	1.00±0.00	
Soybean-millet	0.00	0.00	0.00	0.00	0.22	0.22	1.00±0.00	1.00±0.00	
Total	0.15	0.23	0.06	0.34	0.22	1.00			

Table 3: Accuracy and error matrix in proportion of area of the classified map. * 95% confidence interval.





Journal of Statistical Software

MMMMMM YYYY, Volume VV, Issue II.

doi: 10.18637/jss.v000.i00

dtwSat: Time-Weighted Dynamic Time Warping for satellite image time series analysis in R

Victor Maus
INPE

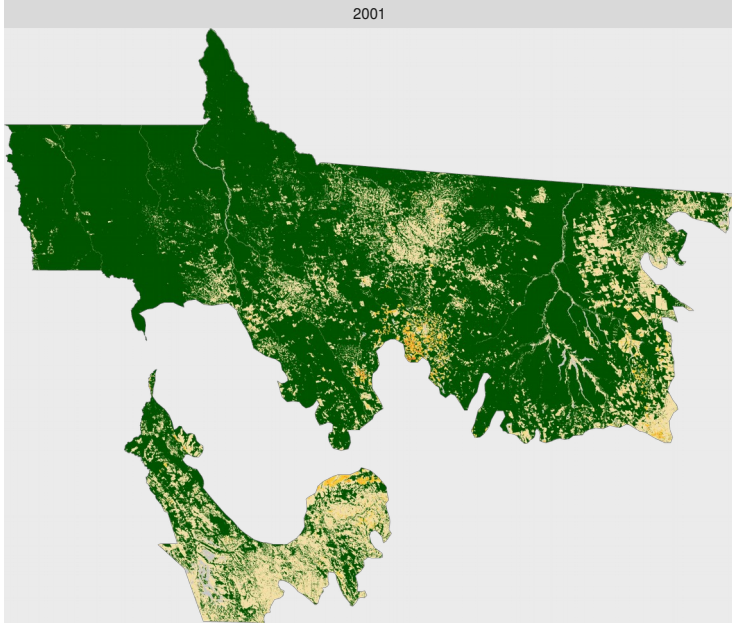
Gilberto Câmara
INPE

Marius Appel
University of Münster

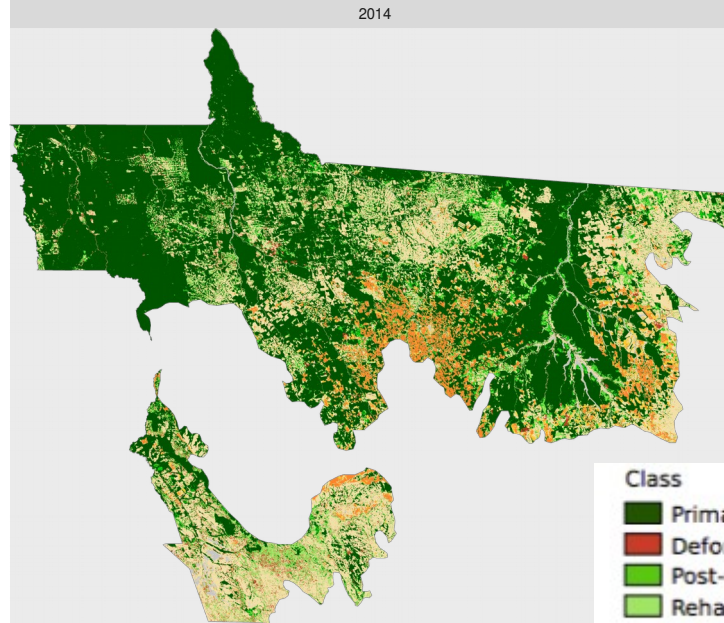
Edzer Pebesma
University of Münster

Larger area - Mato Grosso, Brazil

2001



2014

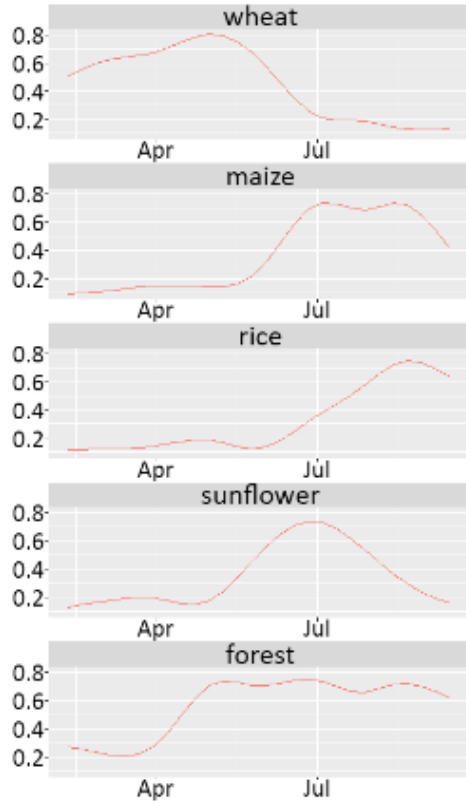


Class

- Primary forest
- Deforestation/Forest degradation
- Post-extraction/fire secondary forest
- Rehabilitated secondary forest
- Pasture
- Cotton
- Soybean
- Soybean-cotton
- Soybean-maize
- Unclassified

OA (%) = 80 - Using TWDTW distance to select the most similar class.

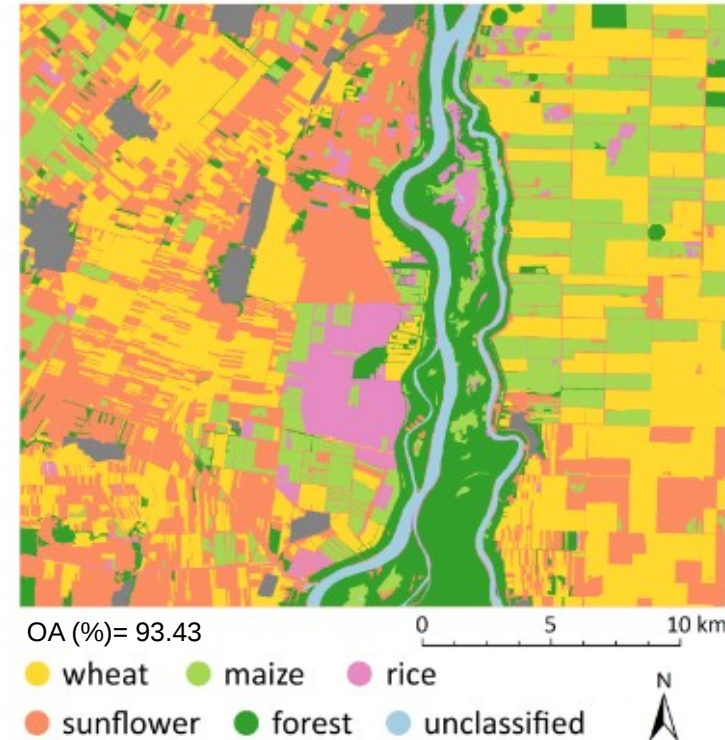
OA (%) = 91 - Using TWDTW distances as input to a SVM model.



Combining TWDTW with object based using Sentinel-2 NDVI time series in Romania.

Open source tools can speed up the development of new approaches.

Figure 4: Object-based TWDTW classification of the study area. The settlements (gray color) are masked.



Thank you!

Victor Maus, maus@iiasa.ac.at

dtwSat on CRAN <https://cran.r-project.org/web/packages/dtwSat/index.html>

dtwSat development on GitHub <https://github.com/vwmaus/dtwSat/>