

# Comparison of the vegetation anomalies products used in the crop monitoring systems

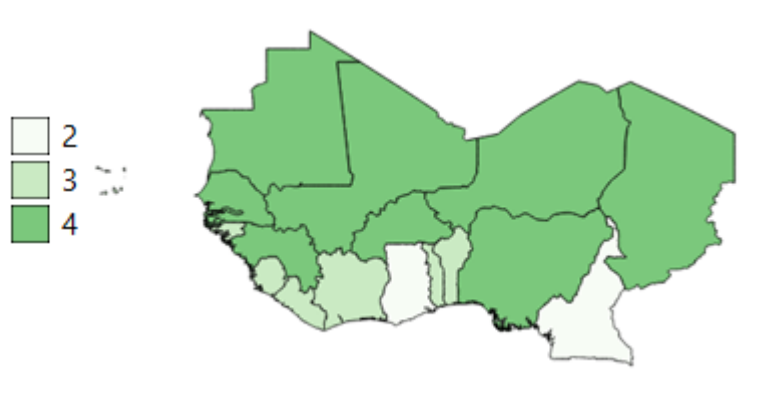
West Africa study case

**Agnès Bégué, L. Lemettais, S. Madec, L. Leroux, R. Interdonato**

# 40 years of Earth Observation for crop monitoring in countries at risk ...



More than 5 Crop Monitoring Systems for food security in West Africa

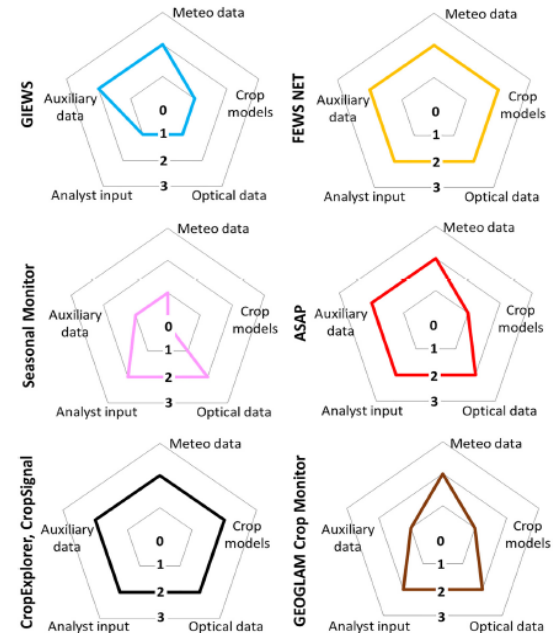
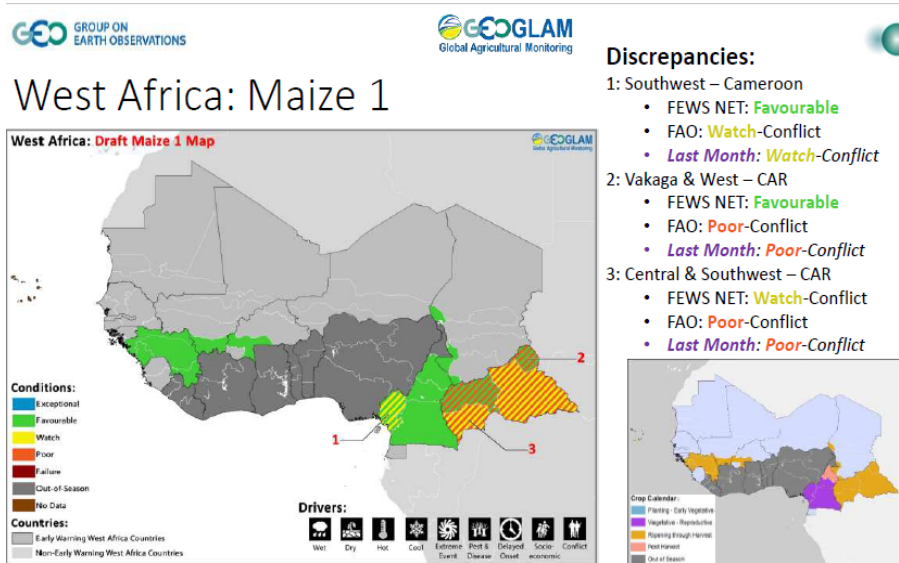


*Number of operational Crop Monitoring systems in West Africa (FEWS -NET, GIEWS, ASAP, VAM, AGRHYMET, CROPWATCH)*



© crédit-photo / Louise Leroux

# ... and still some discrepancies in crop condition assessment



Example of discrepancy map on the maize crop conditions in West Africa, as reported by FEWS NET and GIEWS.

Source: Courtesy of GEOGLAM Crop Monitor  
Becker-Reshef et al., 2020

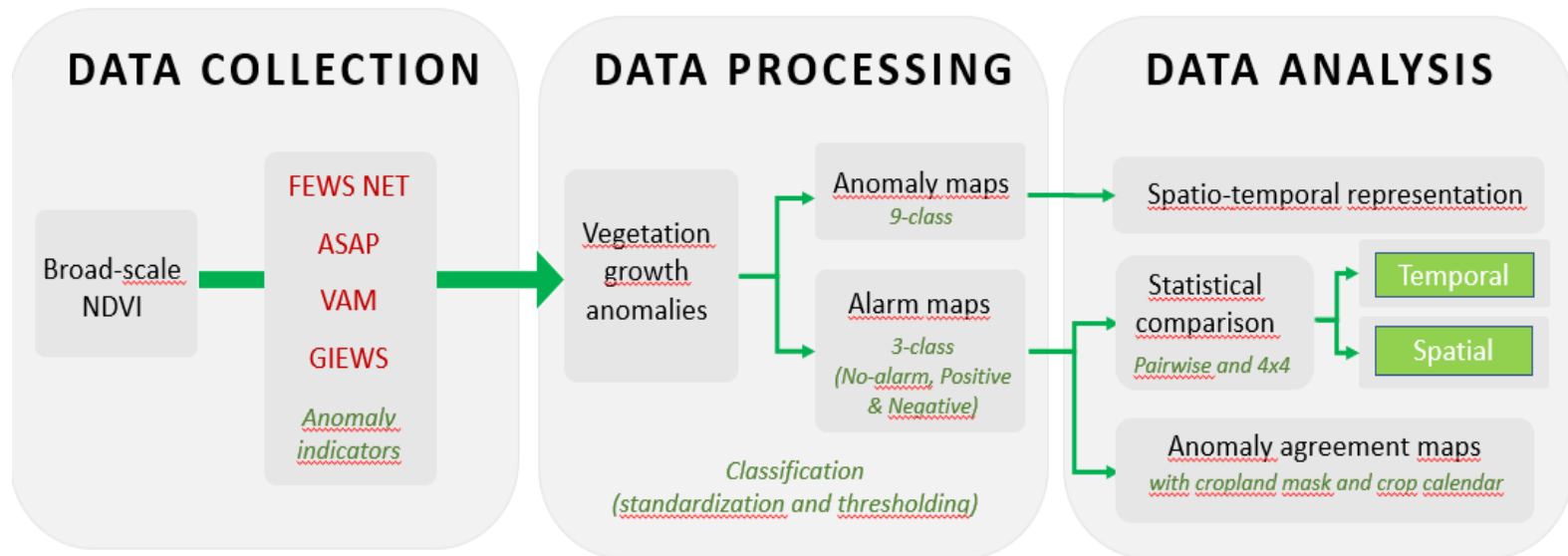
A comparison of global agricultural monitoring systems (sources of input data) and current gaps. Fritz et al., 2019

# What can explain these discrepancies?

A comparative experiment of **growth vegetation anomalies** produced by the main Crop Monitoring Systems in West Africa for the 2010-2020 period

- Are there temporal or spatial patterns of discrepancies?
- What consequences for the Early Warning Systems for food security?
- How to compare different vegetation anomalies both in time and space?

# The approach



A set of four NDVI-based growth anomaly indicators was selected, harmonized, classified and compared in time and space

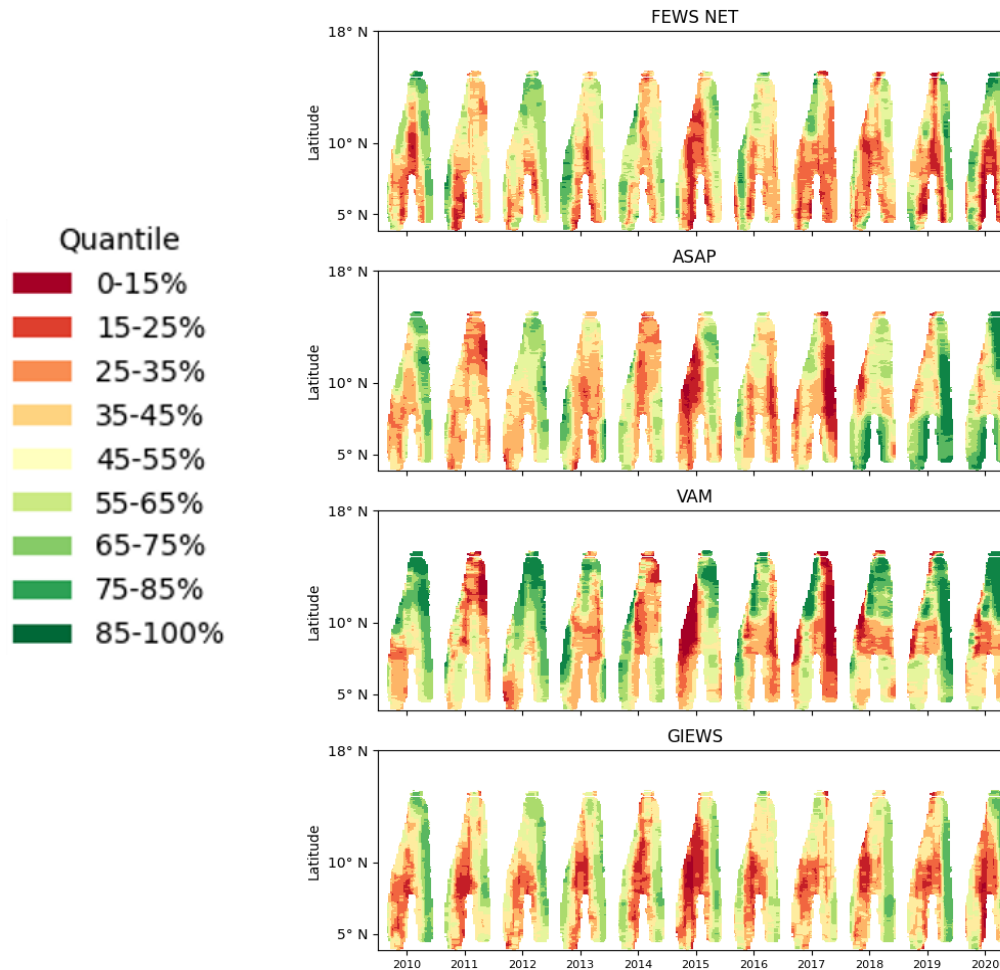
# Data collection



Selection of 4 NDVI-based anomaly indicators from 4 operational crop monitoring systems

	FEWS NET	VAM	ASAP	GIEWS*
Satellite product	eMODIS Level-1B Collection 6 (MODIS Terra & Aqua)	MYD13C1-MOD13C1 (MODIS Terra & Aqua)	MOD13A2-MYD13A2 V006 (MODIS Terra & Aqua)	NOAA-AVHRR & METOP (since 2007)
Pre-processing	Weighted least-squares linear regression smoothing	Whittaker filter	Whittaker filter	Weighted least-squares linear regression smoothing
NDVI-based Anomaly indicator	% median: NDVI	% mean: NDVI	z-score: NDVI	% mean: NDVI
Spatial resolution	250 m	5.6 km	1 km	1 km (since 2007)
Frequency	10 days	8 days	10 days	10 days
Time reference	2003-2017	2002-2013	10/2001 to 12/2020	1984-2014
Web application	Early Warning eXplorer (EWX) <a href="https://earlywarning.usgs.gov/fews/search">https://earlywarning.usgs.gov/fews/search</a>	Hunger Analytics Hub <a href="https://dataviz.vam.wfp.org/Hunger-Analytics-Hub">https://dataviz.vam.wfp.org/Hunger-Analytics-Hub</a>	ASAP Warning Explorer <a href="https://mars.jrc.ec.europa.eu/asap/wexplorer/">https://mars.jrc.ec.europa.eu/asap/wexplorer/</a>	ASIS Global indicators <a href="https://www.fao.org/giews/earthobservation/asis/index_2.jsp?lang=en">https://www.fao.org/giews/earthobservation/asis/index_2.jsp?lang=en</a>

# Vegetation anomalies Hovmöller plots 2010-2020 for West Africa

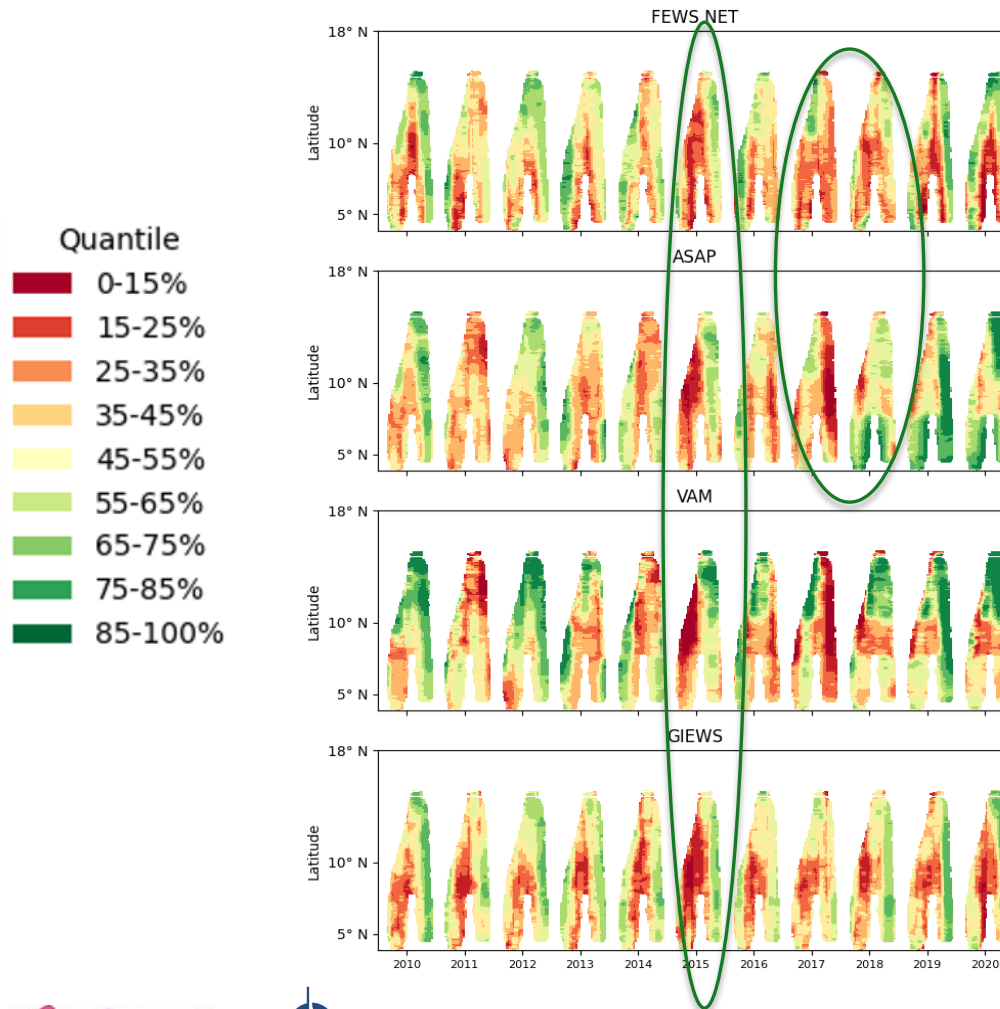


Cropland mask  
GLC-SHARE  
(Latham et al., 2014)

+

Growing season mask  
ASAP Phenological indices  
(Rembold et al., 2019)

# Vegetation anomalies Hovmöller plots 2010-2020 for West Africa




Cropland mask  
GLC-SHARE  
(Latham et al., 2014)


+

Growing season mask  
ASAP Phenological indices  
(Rembold et al., 2019)

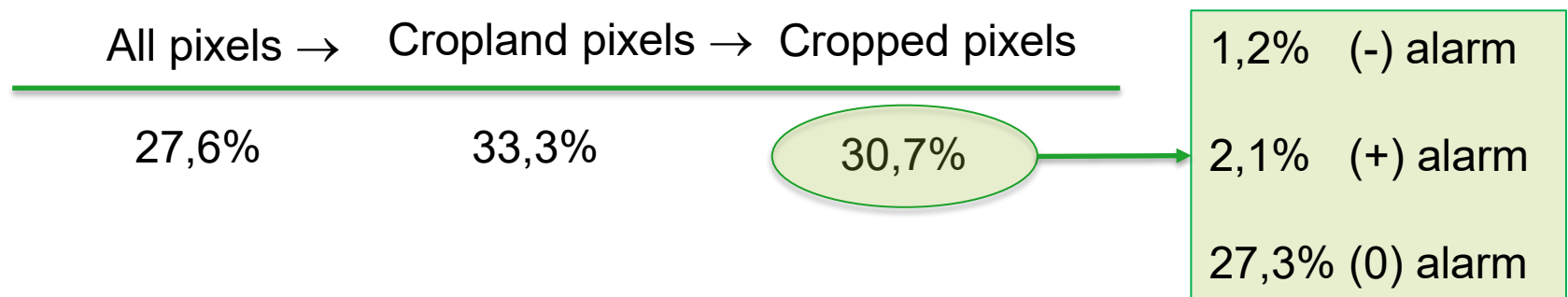


# From vegetation anomalies maps to alarm maps



 To simplify the spatial analysis, the extreme classes corresponding to <15% and >85% of the rank percentile values over the 2010-2020 period were respectively labelled as “negative alarm” and “positive alarm” classes.

 The **alarm maps** of the 4 systems together, and pairwise were then compared.

*Mean of the annual similarities of the 3-class alarm maps of the 4 systems*







# From vegetation anomalies maps to alarm maps

-  To simplify the spatial analysis, the extreme classes corresponding to <15% and >85% of the rank percentile values over the 2010-2020 period were respectively labelled as “negative alarm” and “positive alarm” classes.
-  The alarm maps of the 4 systems together, and pairwise were then compared.

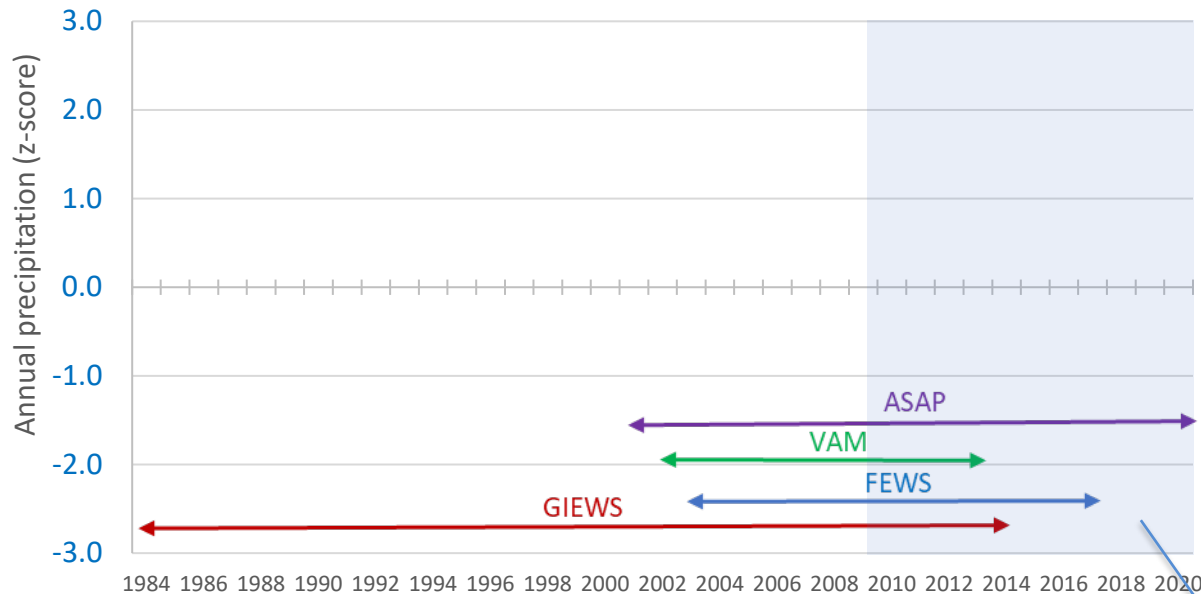
*Pairwise Spearman rank correlation between the 4 systems*

	FEWS NET	VAM	GIEWS	ASAP
FEWS NET	1			
VAM	0.08	1		
GIEWS	-0.01	0.11	1	
ASAP	0.11	0.25	0.06	1

# The potential sources of discrepancies

-  Satellite data (sensor, spatial resolution ...)
-  Satellite time series pre-processing
-  Vegetation anomaly indices : % mean, % median, z-Score
-  Period of reference

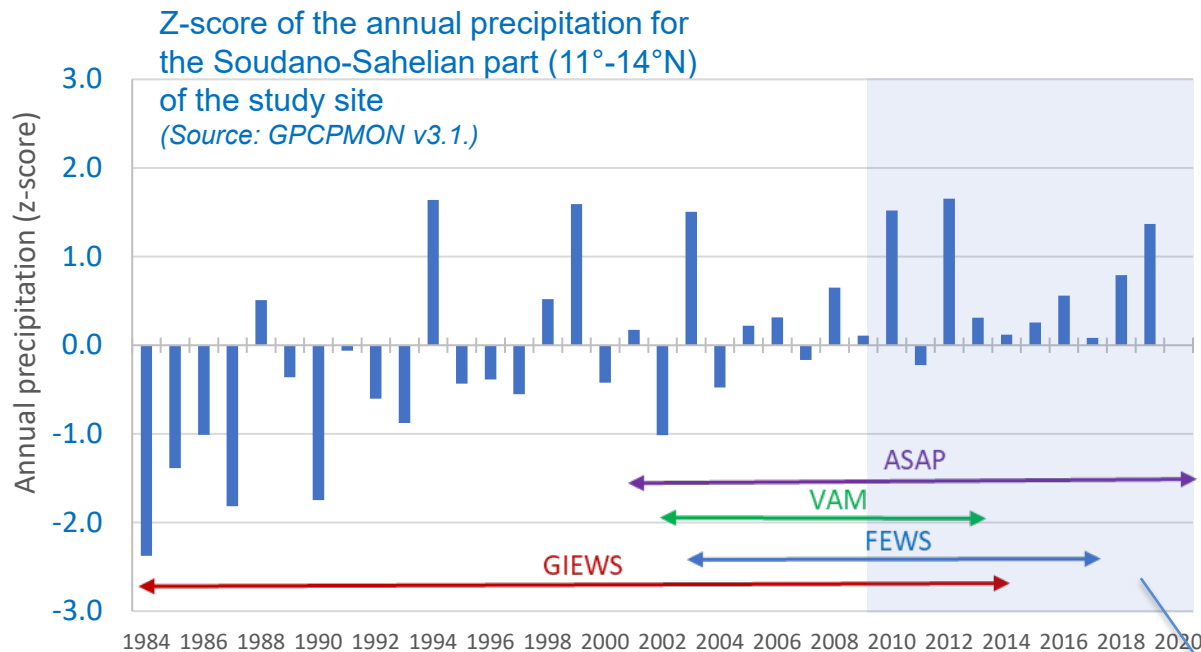
# The potential sources of discrepancies



The colored horizontal lines indicate, for each system, the period of reference used to calculate the vegetation anomalies

The studied period (2010-2020).

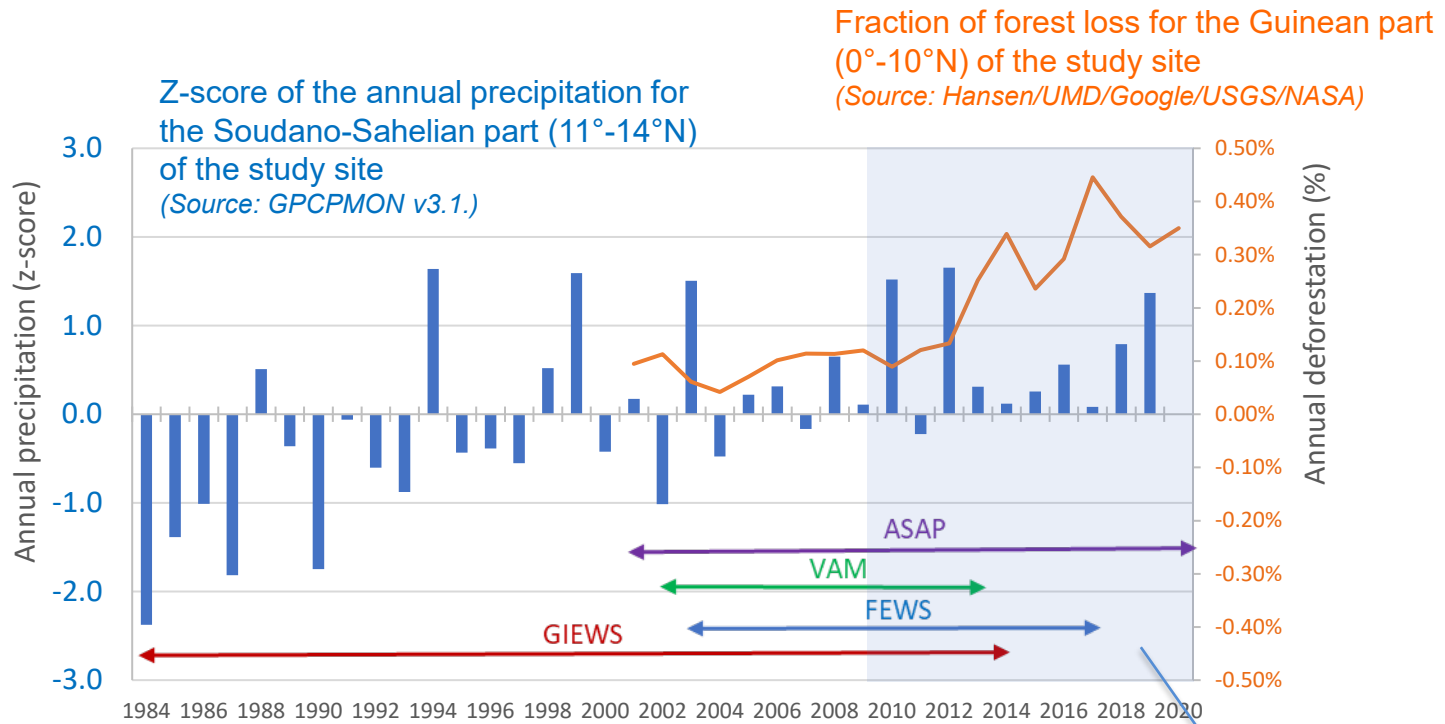
# The potential sources of discrepancies



The colored horizontal lines indicate, for each system, the period of reference used to calculate the vegetation anomalies

The studied period (2010-2020).

# The potential sources of discrepancies



The colored horizontal lines indicate, for each system, the period of reference used to calculate the vegetation anomalies

The studied period (2010-2020).

# What's next ?



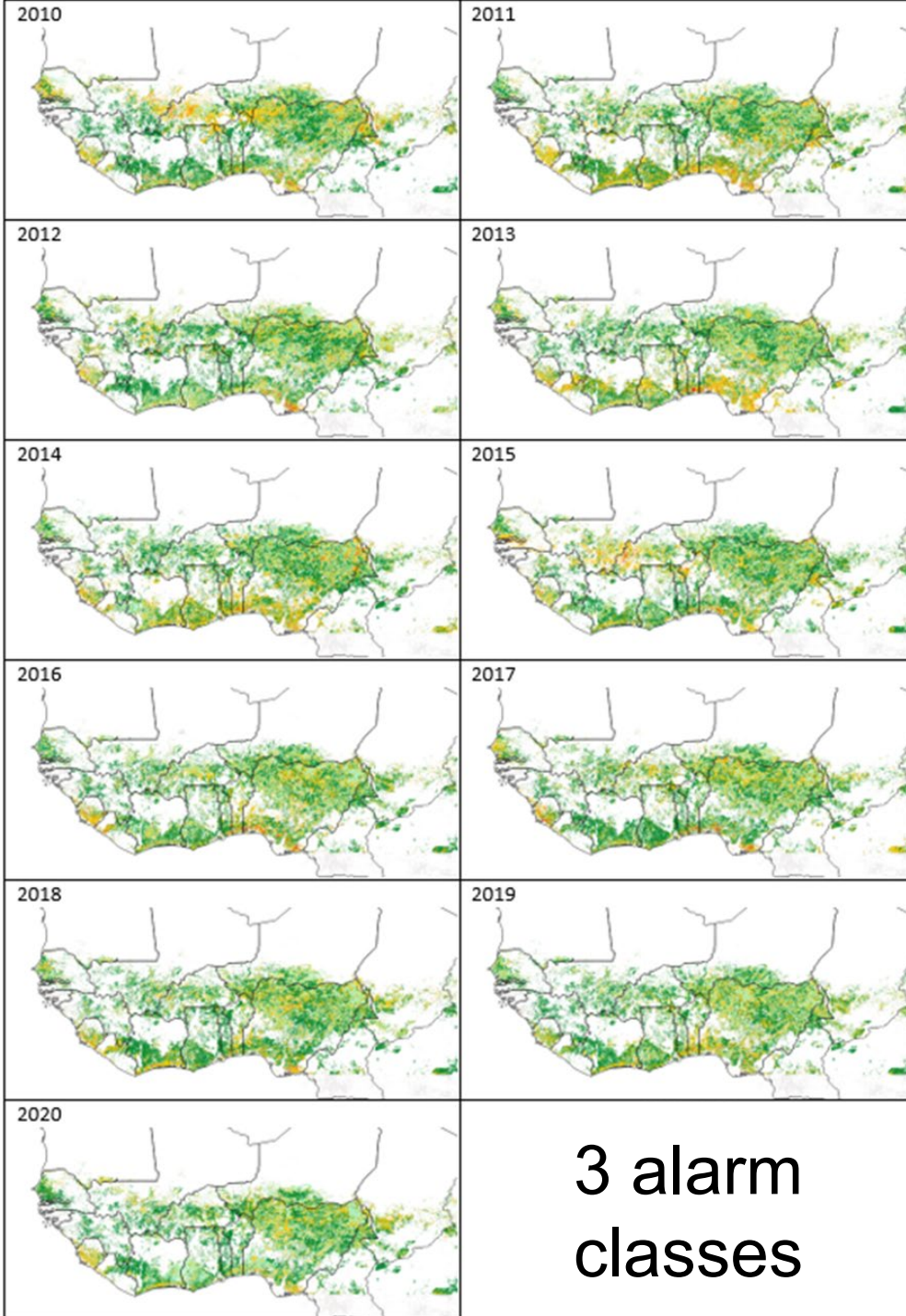
Statistical comparison of the anomaly products

- Unexpected discrepancies between the systems
- Identification of potential sources of discrepancies



The spatio-temporal analysis -> alarm agreement maps (3 classes)

# The alarm agreement maps



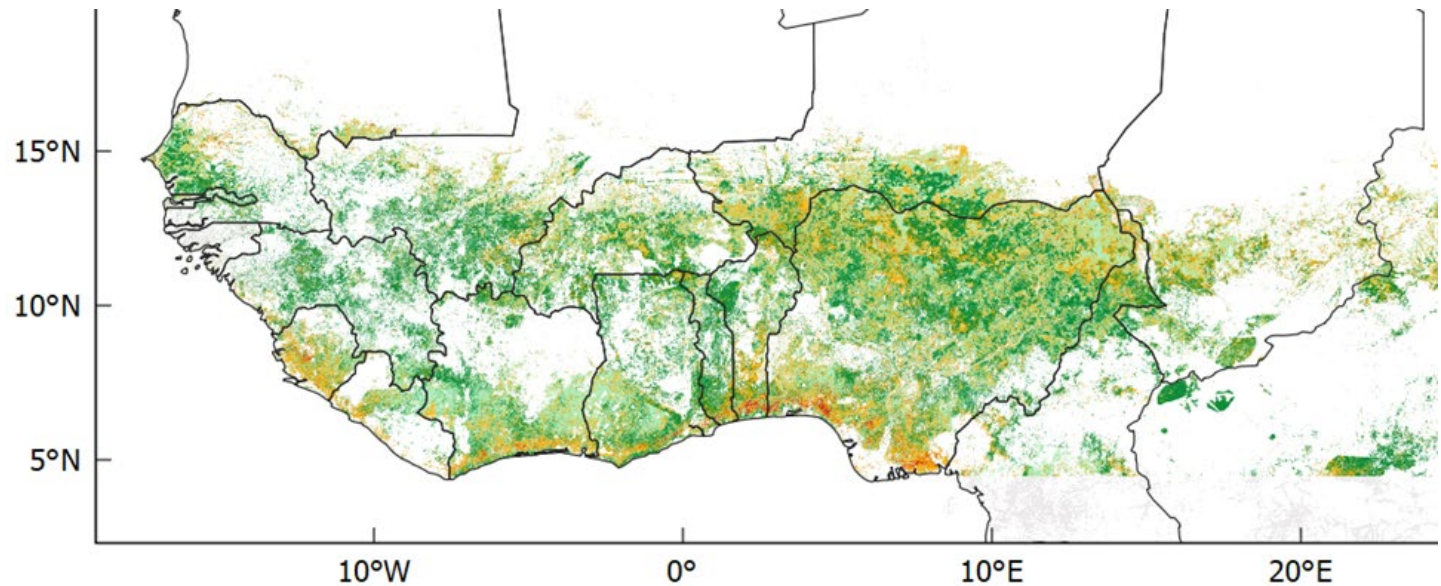
- No agreement (none of the systems have similar classes)
- Low agreement (2 systems have similar classes)
- High agreement (3 systems have similar classes)
- Full agreement (all 4 systems have similar classes)

3 alarm  
classes



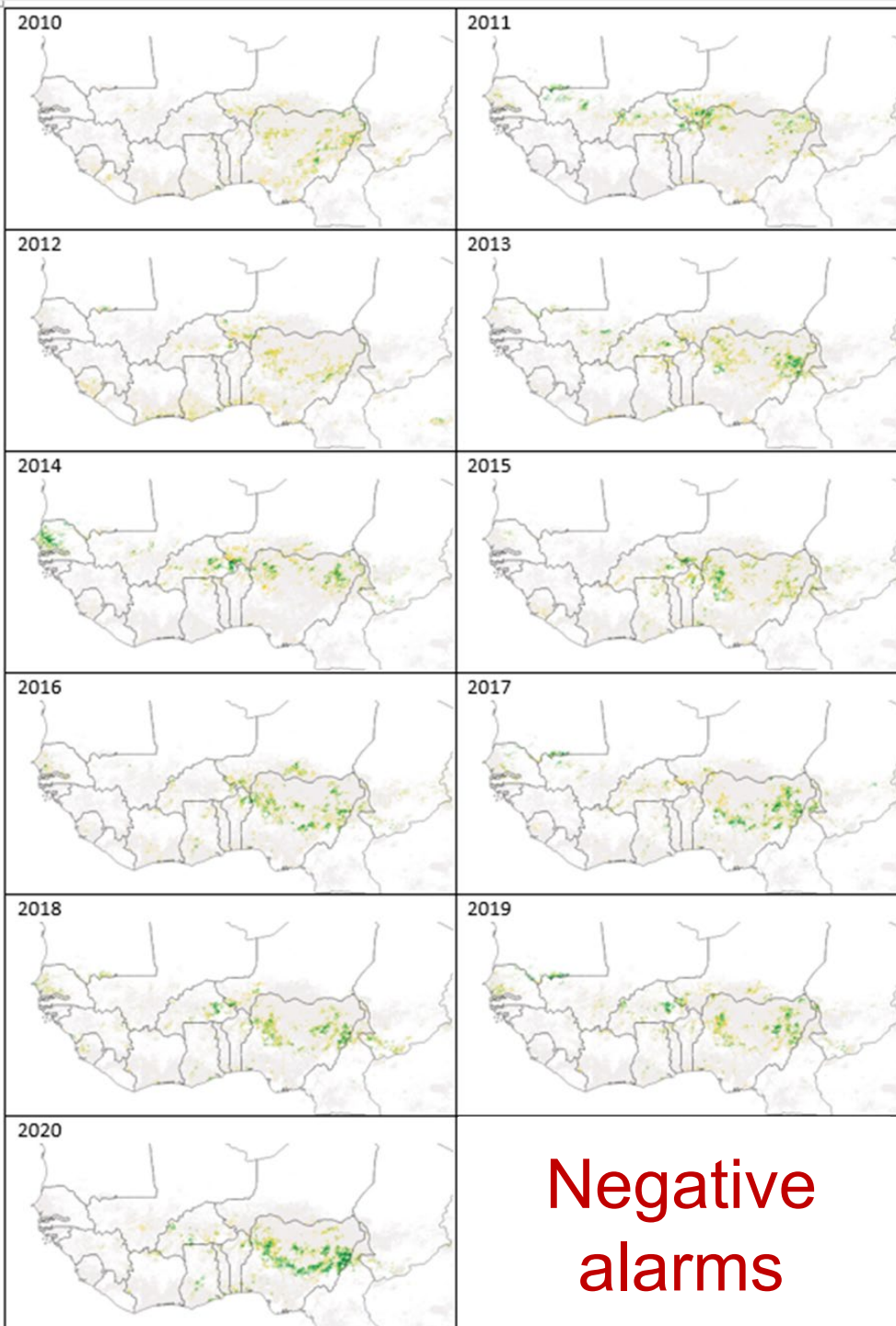
# The alarm agreement maps

*Agreement map of the 3-alarm classes, calculated for the cropland and the crop growing season for the 2010-2020 period*



- No agreement (none of the systems have similar classes)
- Low agreement (2 systems have similar classes)
- High agreement (3 systems have similar classes)
- Full agreement (all 4 systems have similar classes)

# The alarm agreement maps

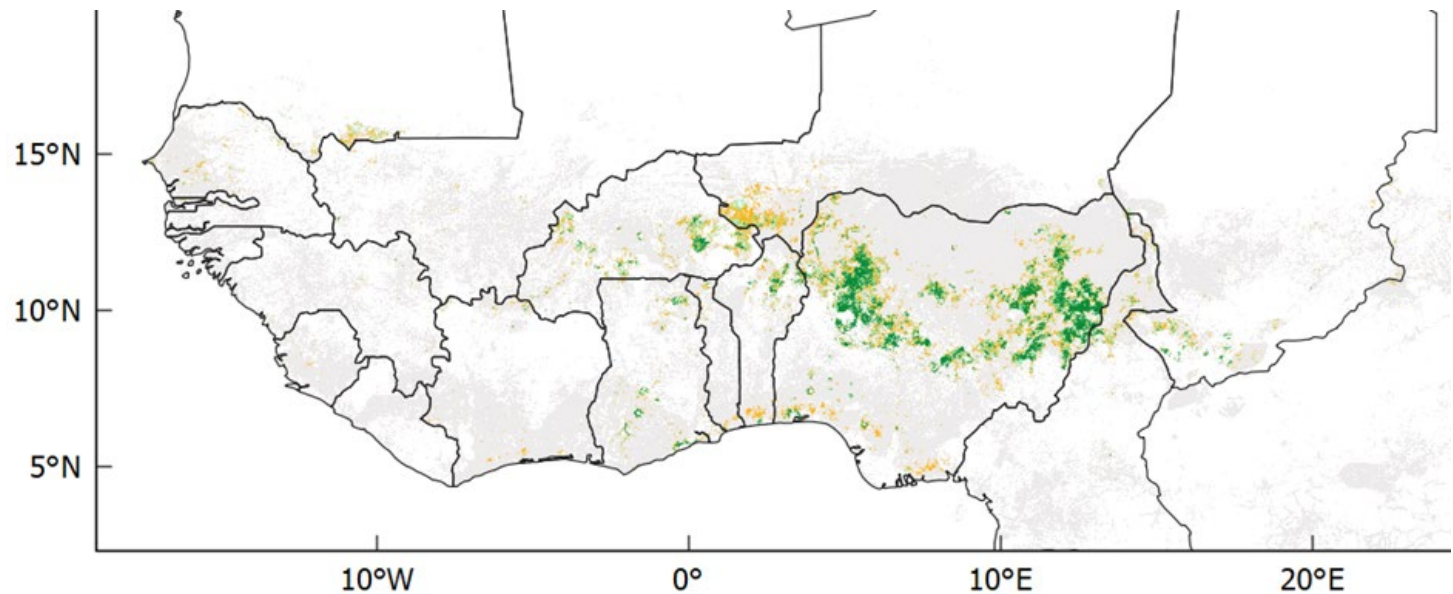


- No agreement (none of the systems have similar classes)
- Low agreement (2 systems have similar classes)
- High agreement (3 systems have similar classes)
- Full agreement (all 4 systems have similar classes)

**Negative  
alarms**

# The alarm agreement maps

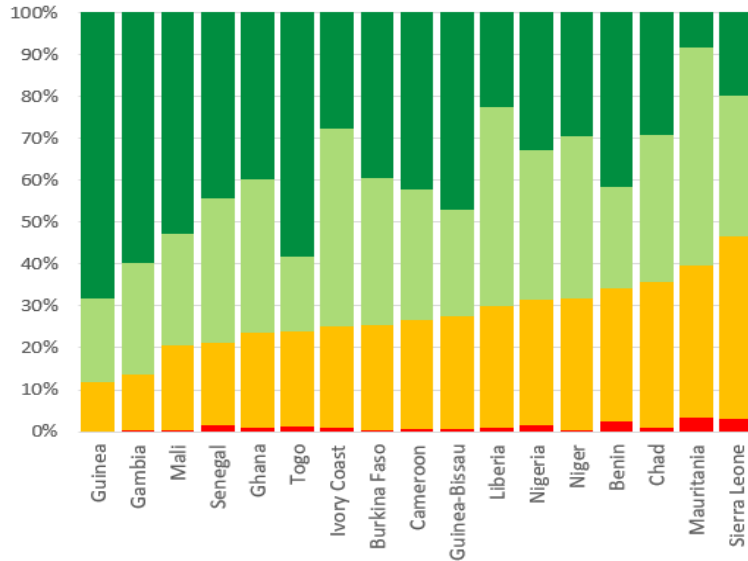
*Agreement map of **negative alarm** class, calculated for the cropland and the crop growing season for the 2010-2020 period*



- No agreement (none of the systems have similar classes)
- Low agreement (2 systems have similar classes)
- High agreement (3 systems have similar classes)
- Full agreement (all 4 systems have similar classes)

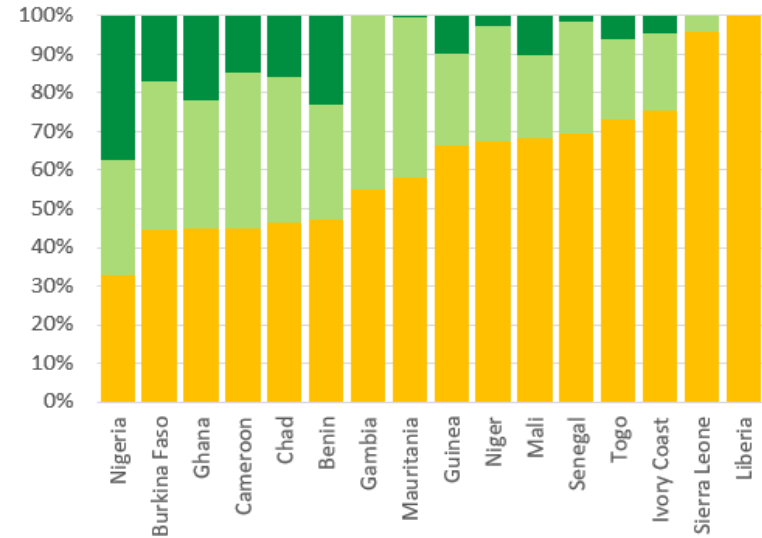
# Geographic analysis

## All 3 alarms






- No agreement (none of the systems have similar classes)
- Low agreement (2 systems have similar classes)
- High agreement (3 systems have similar classes)
- Full agreement (all 4 systems have similar classes)

## Negative alarms



# Conclusions

-  Study limitations :
  - In terms of datasets:
    - Only the NDVI-based anomaly products were considered (other indices exist)
    - Data used for the study (anomaly classes for GIEWS, anomaly values for other systems).
  - In terms of data processing:
    - Potential bias due to the spatial and temporal resampling of the initial products
    - The arbitrary threshold of 15% of the extreme percentiles to define the alarm classes
    - The use of a unique cropland map and of a unique growing season calendar
-  Study results :
  - An approach for spatio-temporal comparison,  
... in the current environment where more and more products are emerging
  - A light on an unexpected source of discrepancies between systems
-  Study promises :
  - Which product to use,  
... in an environment where an increasing number of products are available
  - The negative alarm agreement maps could provide information on the confidence level associated with the negative anomaly -> Early warning system

# NIGER

An aerial photograph of a village in Niger, showing a cluster of small, rectangular buildings with flat roofs, surrounded by a dry, orange-brown landscape. The buildings are arranged in a somewhat circular pattern, with a few trees scattered throughout. The terrain is arid and shows signs of erosion, with some small pits and ridges visible. The overall scene is one of a rural, semi-arid environment.

Many thanks to:  
GEOGLAM-CM4EW (I. Becker-Reshef, C. Justine, B. Barker)  
JRC (H. Kerdiles, M. Meroni)  
WFP (R. Bonifacio)

Maurizio ASCANI