

# living planet symposium | BONN

23–27 May  
2022

TAKING THE PULSE  
OF OUR PLANET FROM SPACE



## ASSESSING CROPLAND ABANDONMENT FROM VIOLENT CONFLICT IN THE SAHEL WITH SENTINEL-2 & GOOGLE EARTH ENGINE

Laure Boudinaud & Alex Orenstein

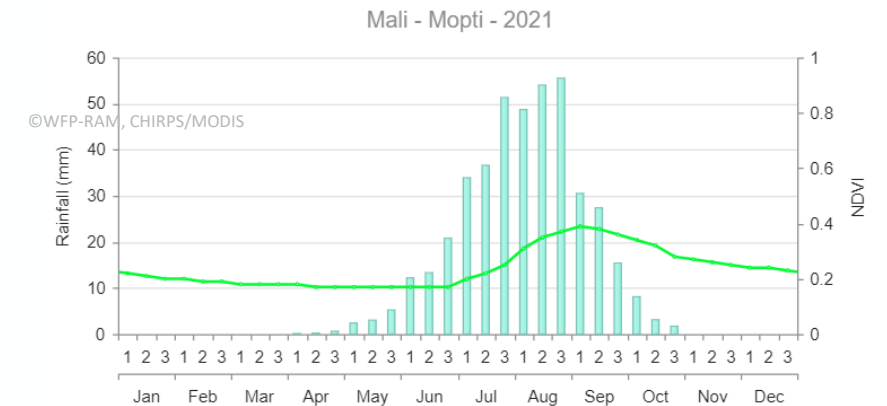
date



- Single rainy season with harvest in Sept/Oct
- Ongoing conflict and humanitarian emergency since 2011 - millions of people food insecure in 2021\*
- Food assistance estimates rely on in-person agricultural surveys impossible during active conflict



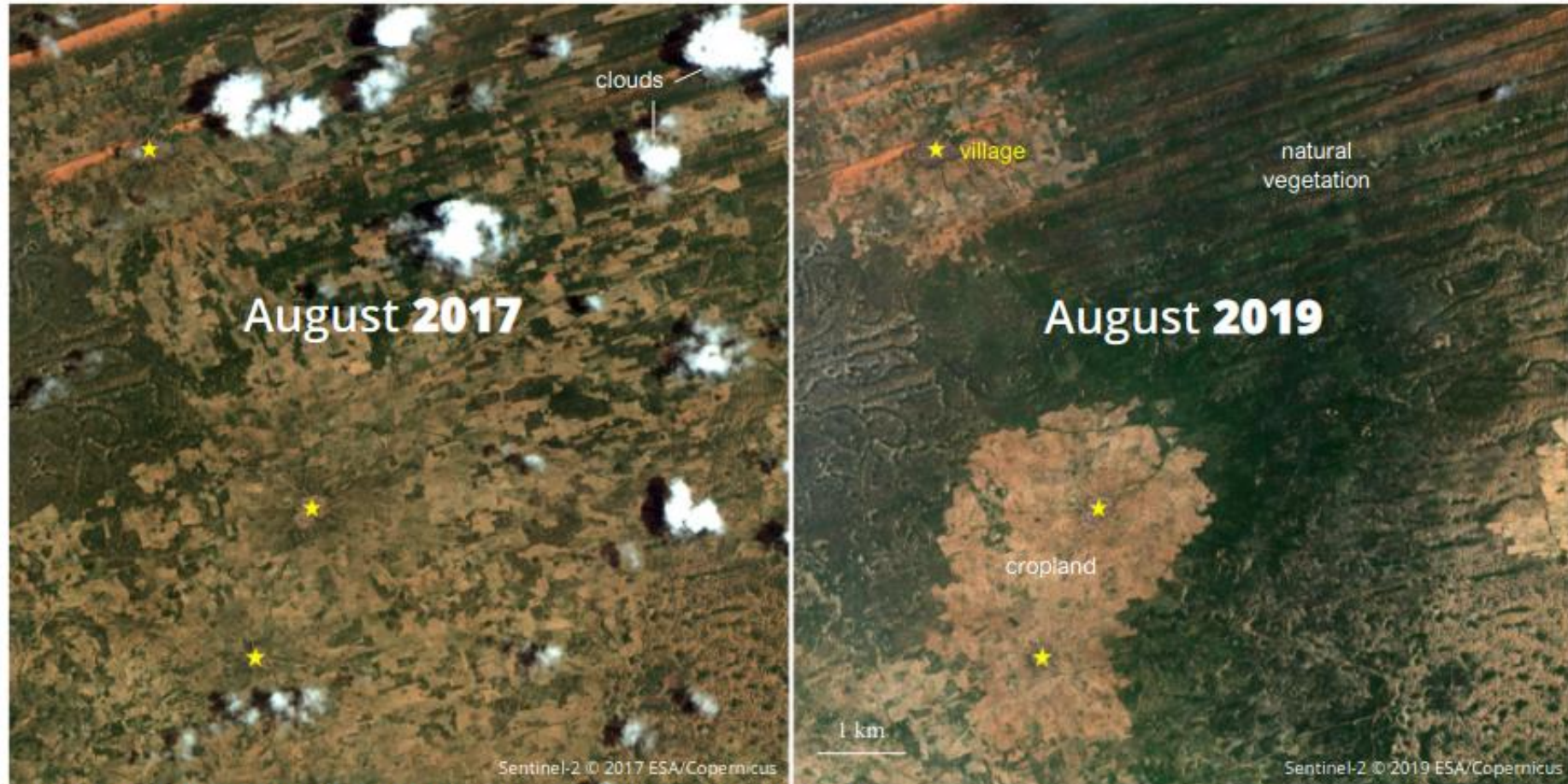
Rainfall



\* Source : Cadre Harmonisé (2021)



# Cropland Abandonment



**Image 1. Cropland abandonment detected from space for three villages in Koro cercle**

In 2017 (left), cultivated fields are visible (in beige) as far as 10 kilometres away from habitations (star), singling out from the surrounding natural vegetation in dark green. In 2019 (right), cultivated fields are considerably reduced and concentrated in proximity to habitations due to movement restrictions imposed by armed actors.

# Overview

- New method to analyse inter-annual cropland change;
  - Easy identification of **cropland abandonment**: assist humanitarian action/ decision-making in hard to reach areas of the Sahel
  - thousands of villages with cropland losses, most of which were **not identified by other humanitarian actors** or early warning systems.
- 3-period timescan:
  - S2-derived product reflecting vegetation evolution during a full growing season,
  - yearly composite of max-NDVI values (beginning, middle and end of growing season),
  - stacked into one RGB image used to **single out cultivated land for each season**.
- Target audience / end users are humanitarian actors;

# Methodology - data & tools



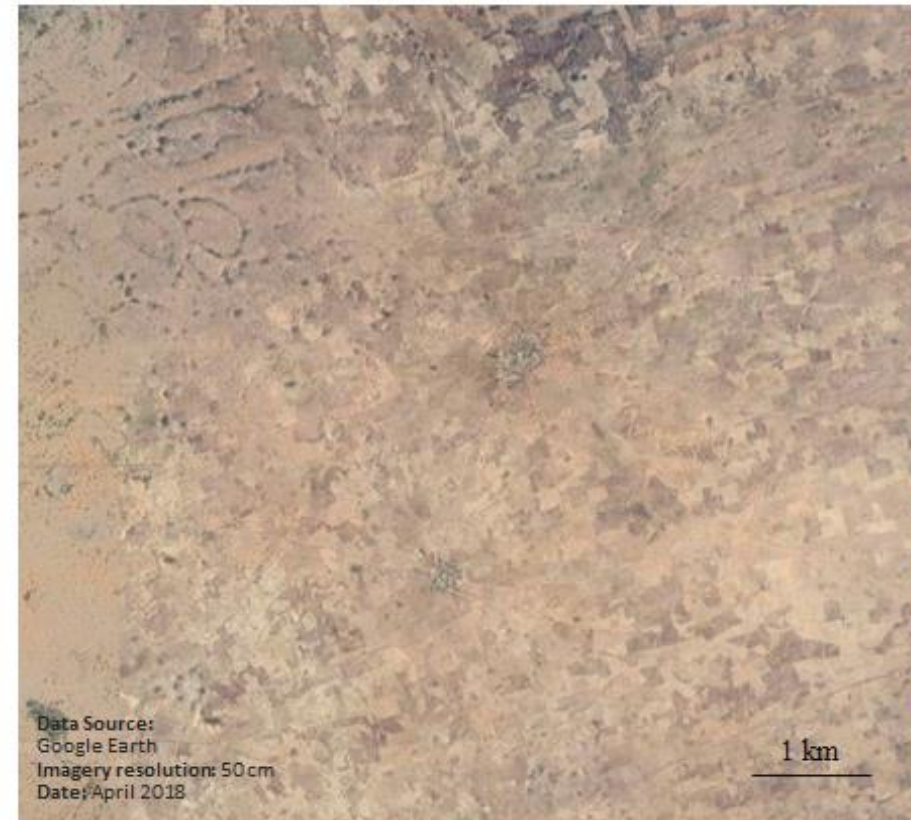
- Sentinel-2
  - Combines most adequate characteristics to detect agricultural fields over the AOI.
- Google Earth Engine
  - Given the amount of S2 images processed, the most appropriate tool to run analyses.

```
29 };
30 // FUNCTION to create new bands to the collection
31 var GREEN_s = 'B3';
32 var RED_s = 'B4';
33 var NIR_s = 'B8';
34 var SWIR_s = 'B11';
35
36 var addIndices = function(image) {
37   var ndvi = image.normalizedDifference([NIR_s, RED_s]);
38   return image.addBands(ndvi.rename('NDVI'));
39 };
40 //var s2016 = sentinel2(2016).map(addIndices);
41 //print('Sentinel collection 2016', s2016);
42 var s2017 = sentinel2(2017).map(addIndices);
43 print('Sentinel collection 2017', s2017);
44 var s2018 = sentinel2(2018).map(addIndices);
45 print('Sentinel collection 2018', s2018);
46 var s2019 = sentinel2(2019).map(addIndices);
47 print('Sentinel collection 2019', s2019);
48 var s2020 = sentinel2(2020).map(addIndices);
49 print('Sentinel collection 2019', s2020);
50
51 // 3-PERIODS PRODUCTS
52 // Visualize RGB from different max-NDVI values from three distinct periods of the year
53 var p1 = s2017.filter(ee.Filter.dayOfYear(t1, t2)).select('NDVI').max();
54 var p2 = s2017.filter(ee.Filter.dayOfYear(t2+1, t3)).select('NDVI').max();
55 var p3 = s2017.filter(ee.Filter.dayOfYear(t3+1, t4)).select('NDVI').max();
56
57 var min_values = p2.reduceRegion('min', region, 100).values().getInfo();
```

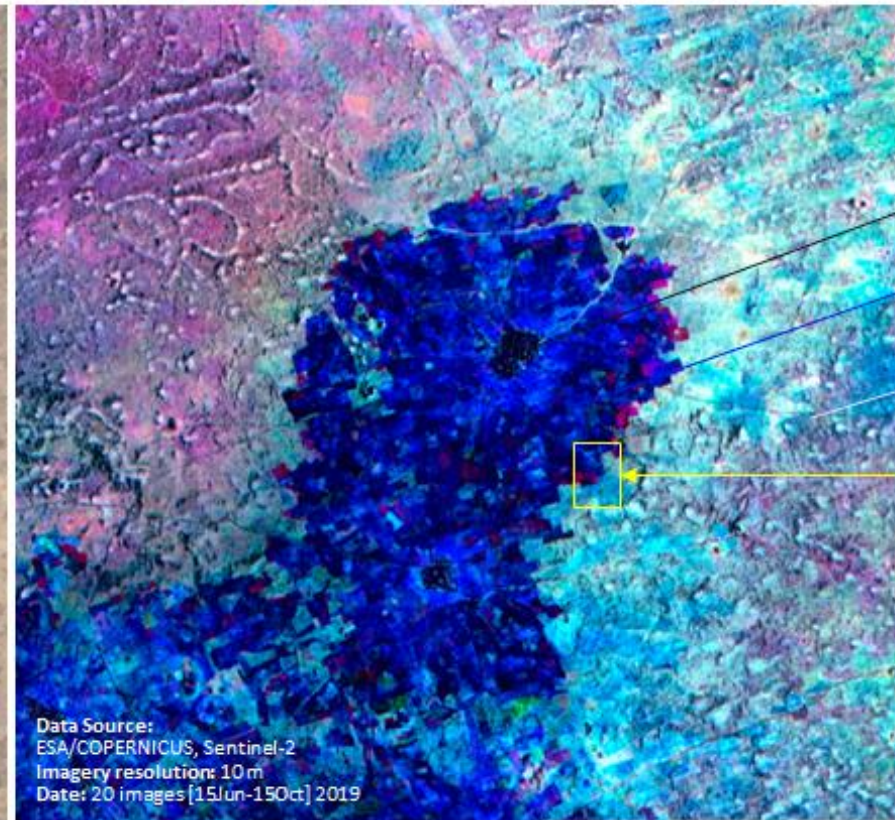


# Methodology - 3-period timescan

Image available on Google Earth



Composite image (timescan) derived from Sentinel-2



Village

Cropland

Natural  
vegetation

Field validation



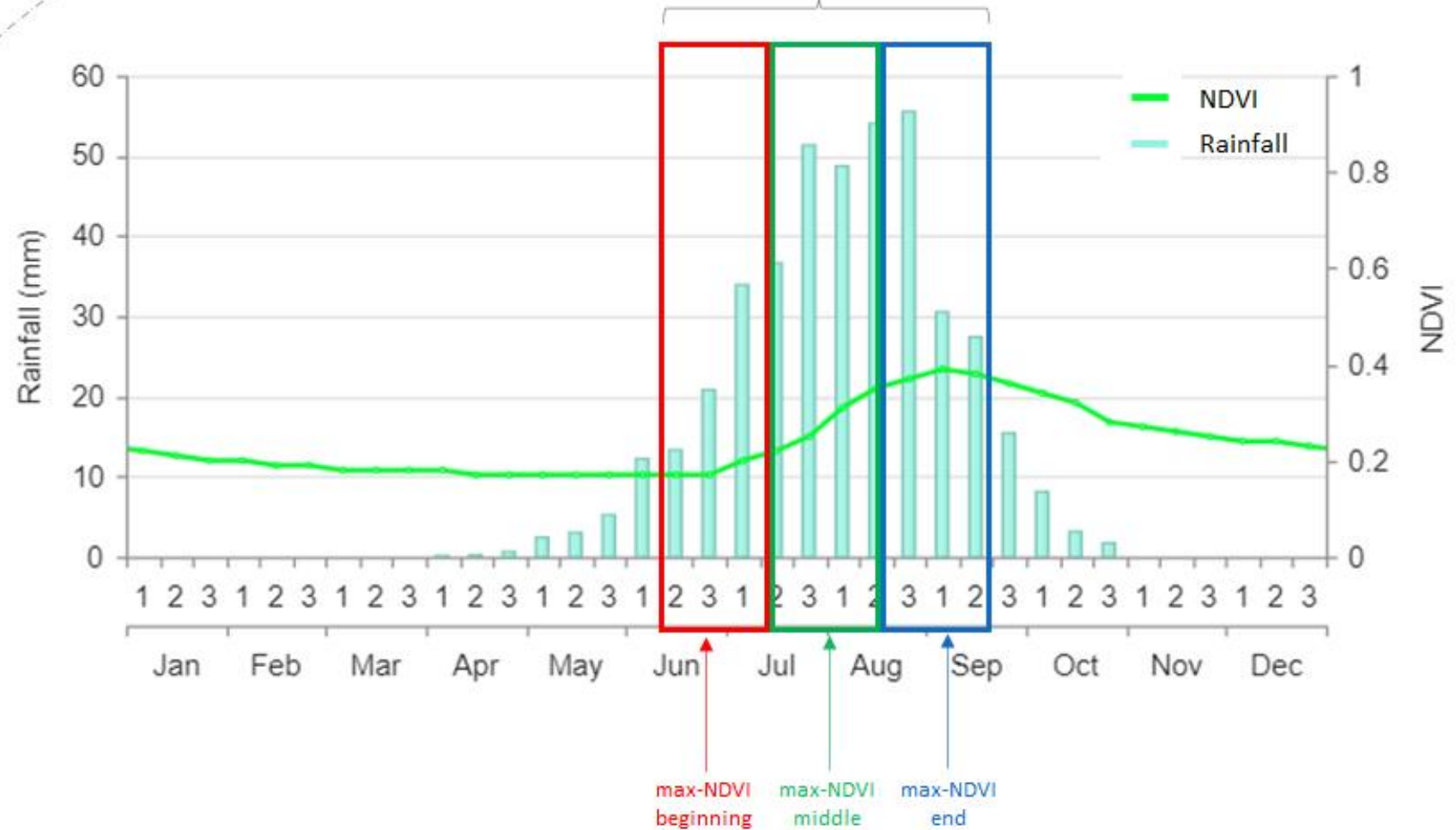
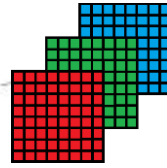
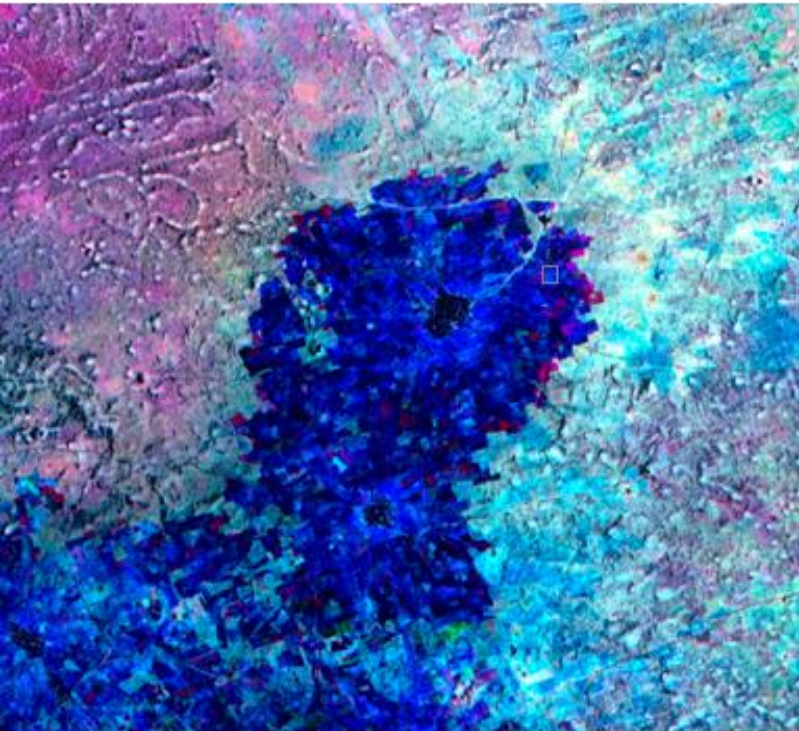
A. Single-date satellite image (source : Google Earth)

B. 3-period timescan : RGB composite derived from imagery acquired between 15 Jun and 15 Oct



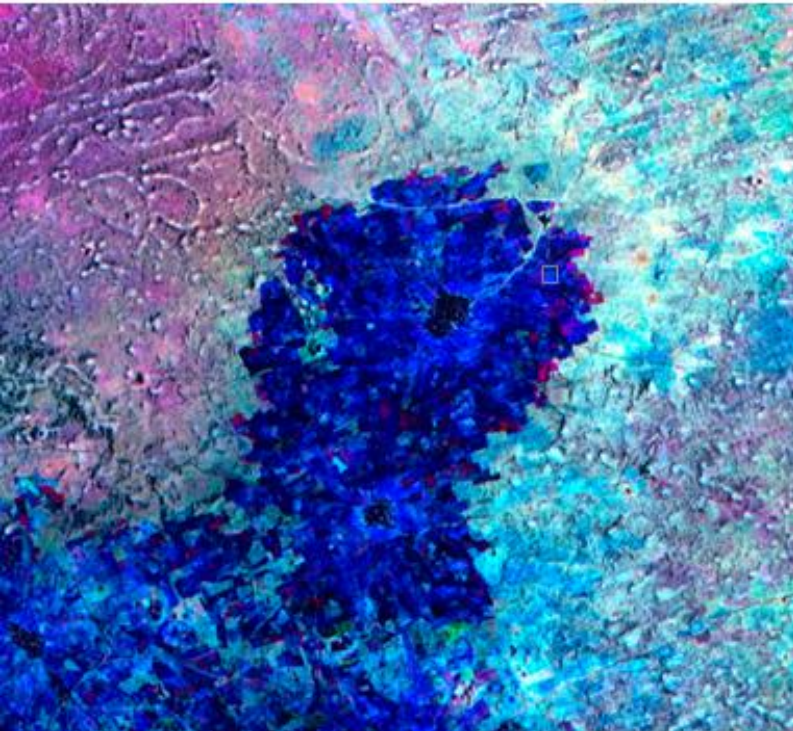
# Methodology

**3-period timescan** = a **RGB** composite of NDVI values along the agricultural season

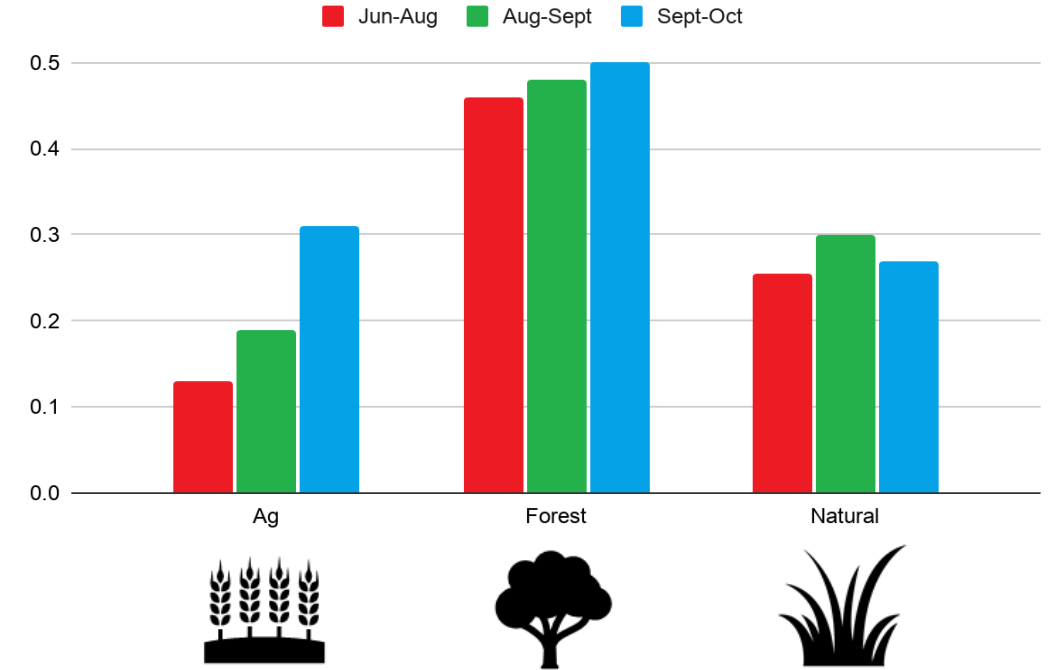


# Methodology













Different land cover types are associated with specific colours, revealing patterns that are easy to interpret



Ag, Forest and Natural



## RGB Color Mixing

	+		+		=	
21		57		148		(21, 57, 148)
	+		+		=	
255		255		255		(255, 255, 255)
	+		+		=	
78		100		84		(78, 100, 84)

Ag

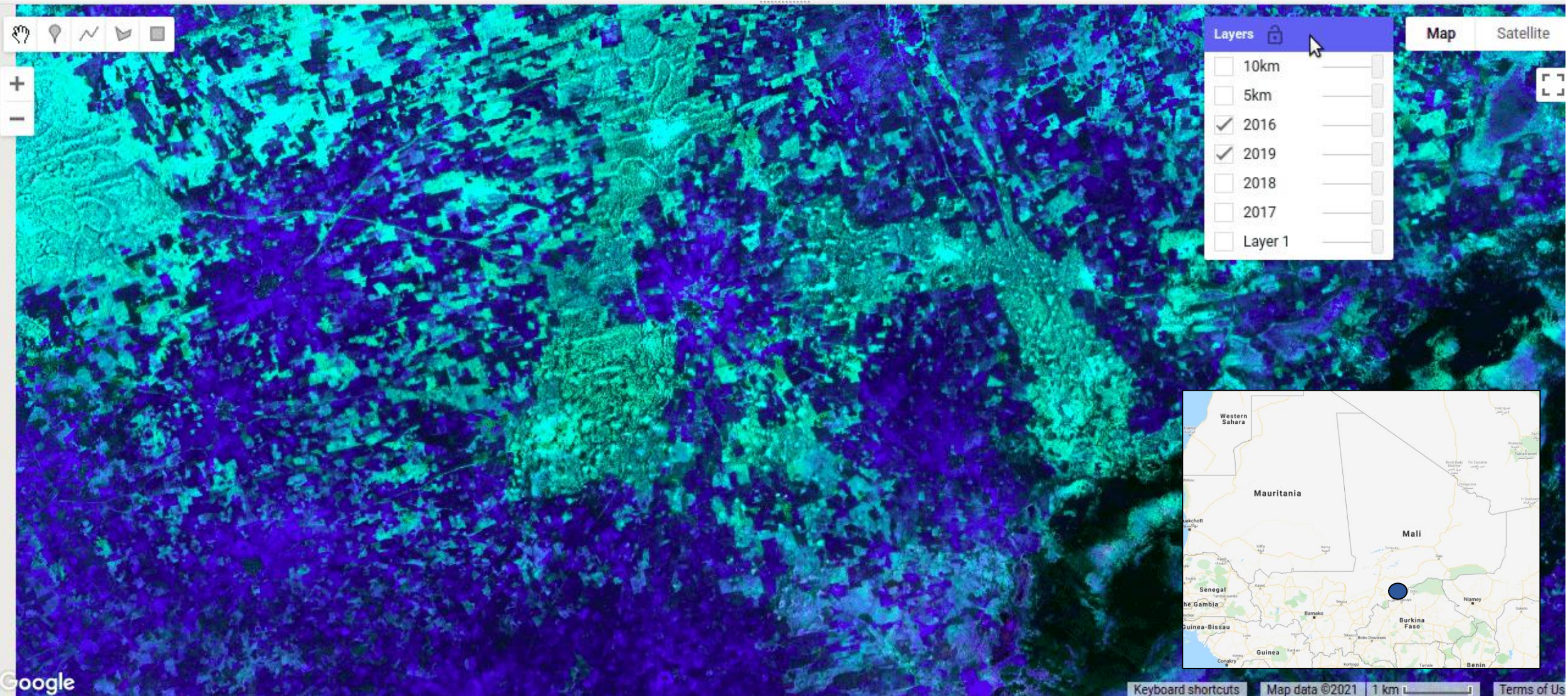
Forest

Natural



<https://tinyurl.com/timescanviz>

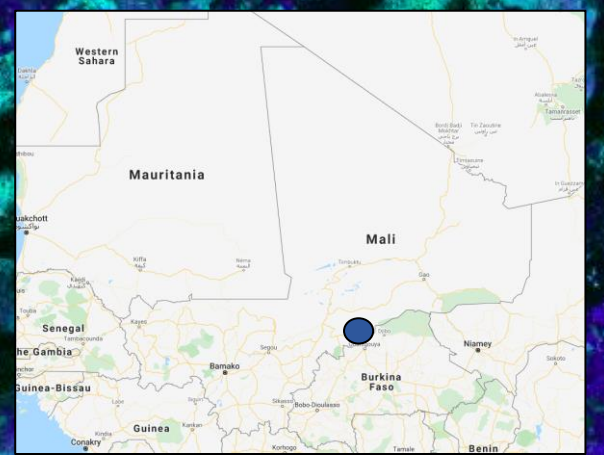
Search places and datasets...



**Layers**

- 10km
- 5km
- 2016
- 2019
- 2018
- 2017
- Layer 1

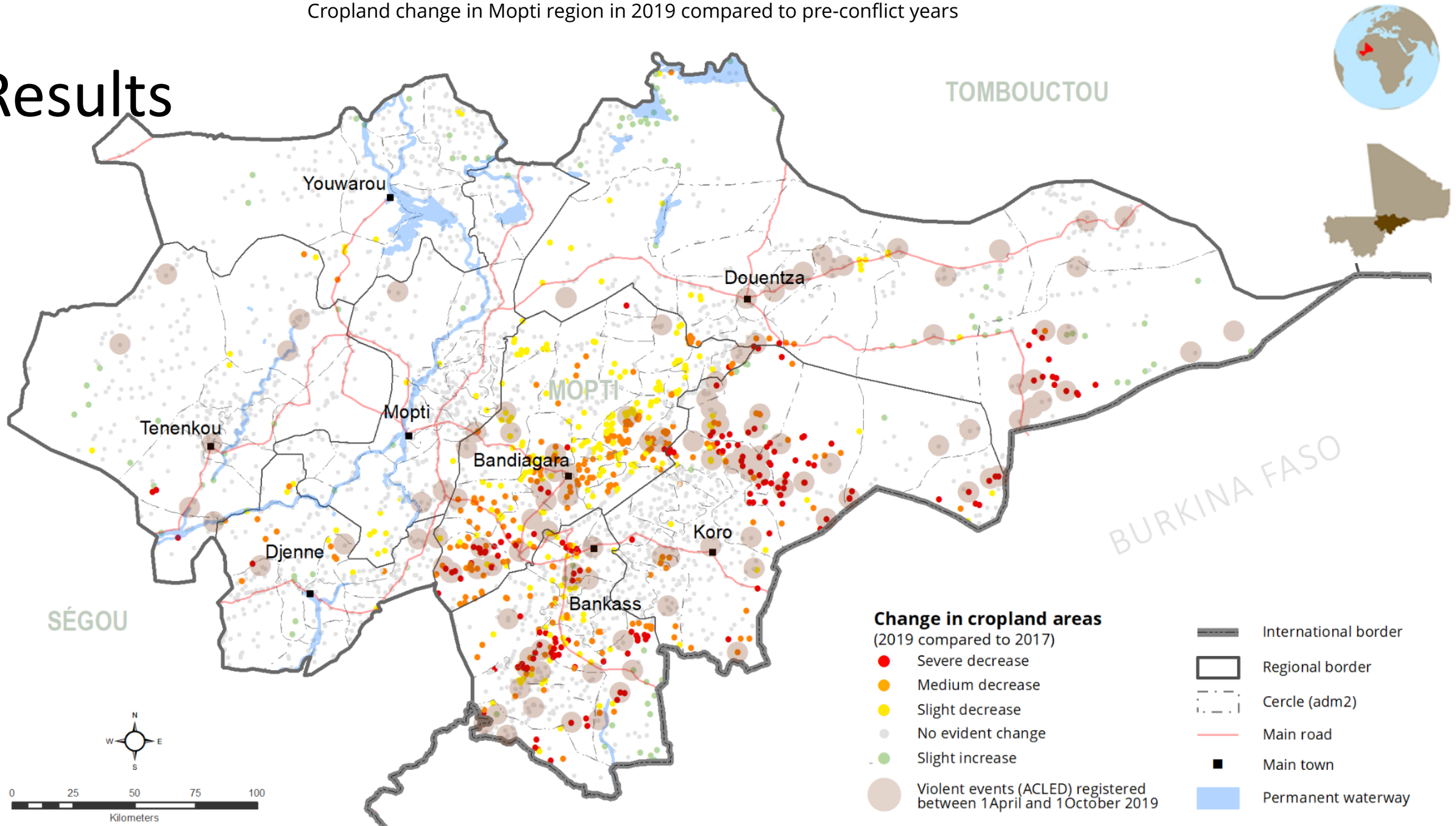
Map Satellite





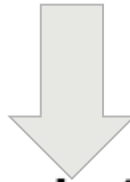
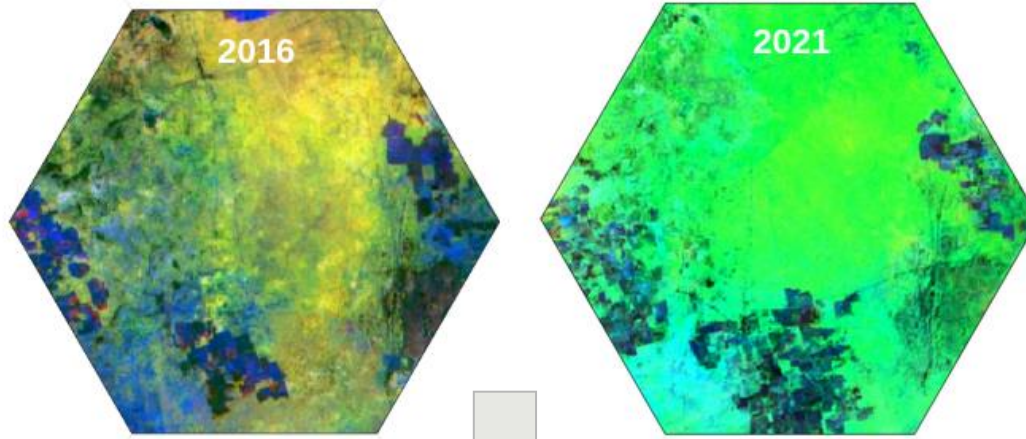
Cropland change in Mopti region in 2019 compared to pre-conflict years

# Results

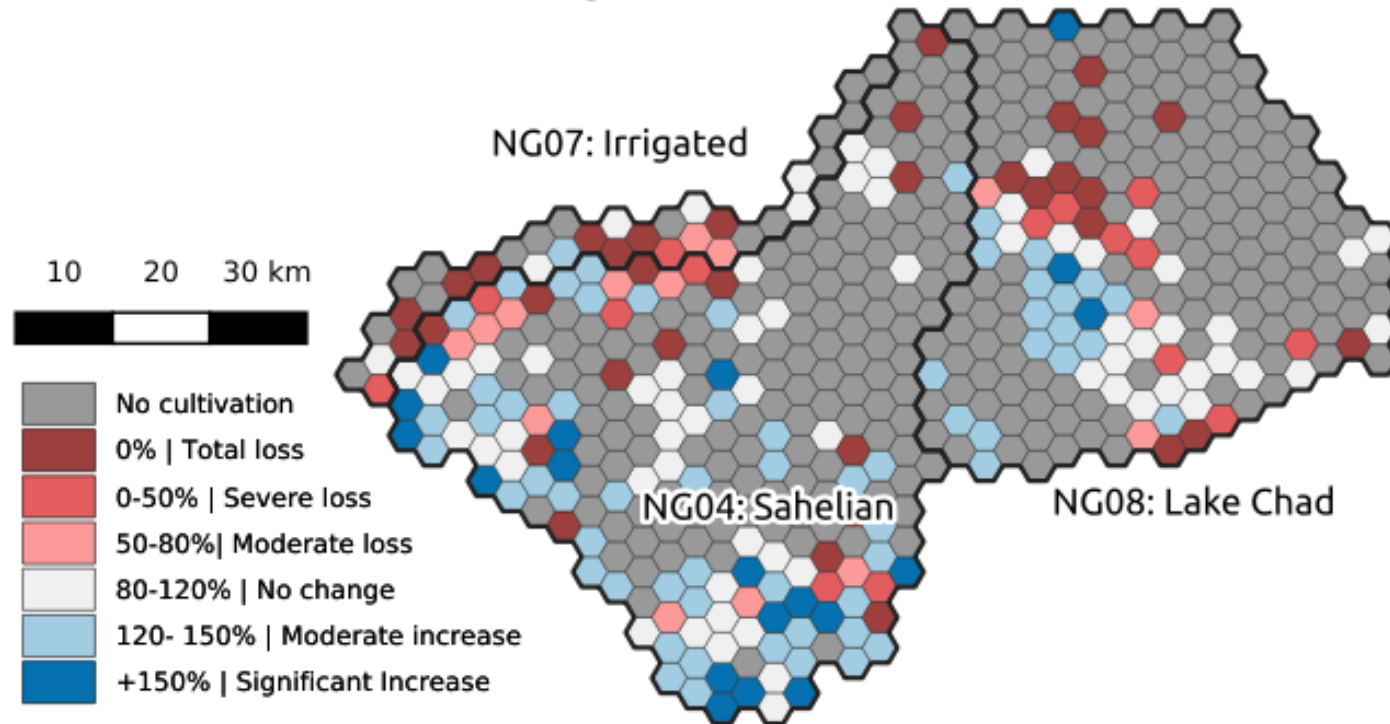




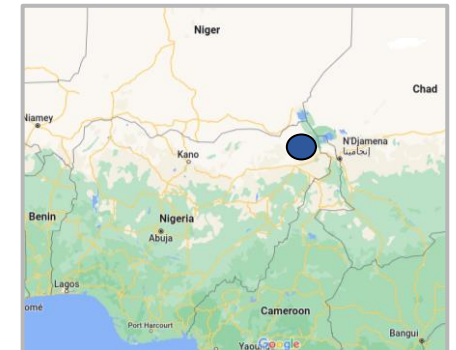
# Results



**Cultivated Surface Changes in Abadam 2016-2021**

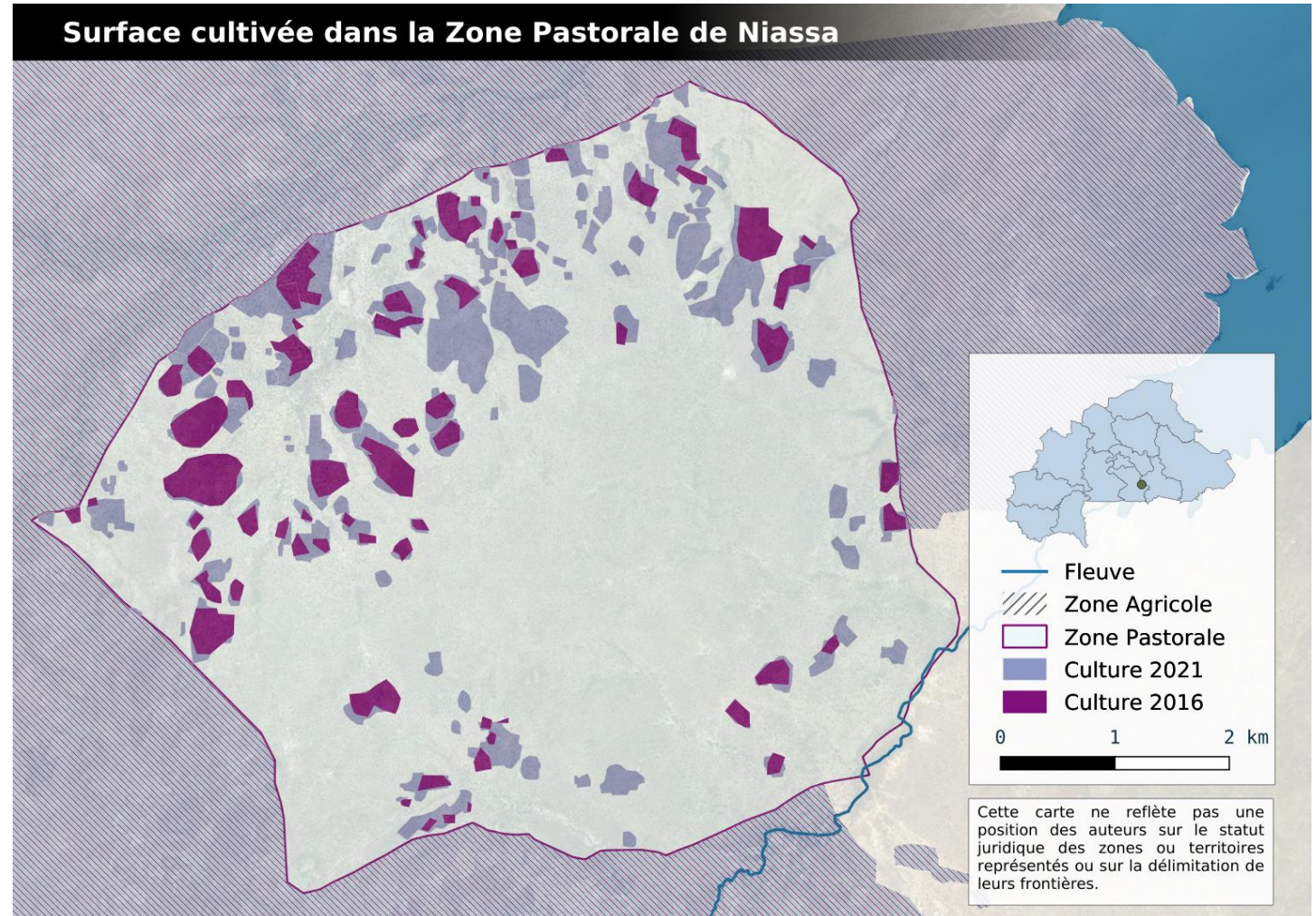


Abadam		Abadam	
		Affected Pop	Zones
No cultivation	NA	42%	58%
Total loss	0%	16%	7%
Severe loss	0 - 50%	4%	3%
Moderate loss	50 - 80%	5%	3%
No change	80 - 120%	20%	13%
Moderate increase	120 - 150%	8%	12%
Significant increase	>150 %	5%	4%



# Operational Uses

- Integrated into **national food security analysis** - to inform in hard-to-reach areas where field surveys cannot be conducted;
- Currently used by humanitarian actors for **geotargeting assistance** in their early response planning.
- Monitoring **agricultural encroachment** into protected areas
- **Validation data** for landcover analyses





# Try it yourself !

Github (GEE + PyQGIS) <https://github.com/oren-sa/3PTS>

GEE App (No code needed)

<https://tinyurl.com/timescanviz>

# Thank you !

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