



Seeing the Gallery Forest for the Savanna Chimpanzee

Remote Sensing for Great Ape Conservation in Kedougou, Senegal

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Pennsylvania State University

MGIS Capstone Project Spring 2024

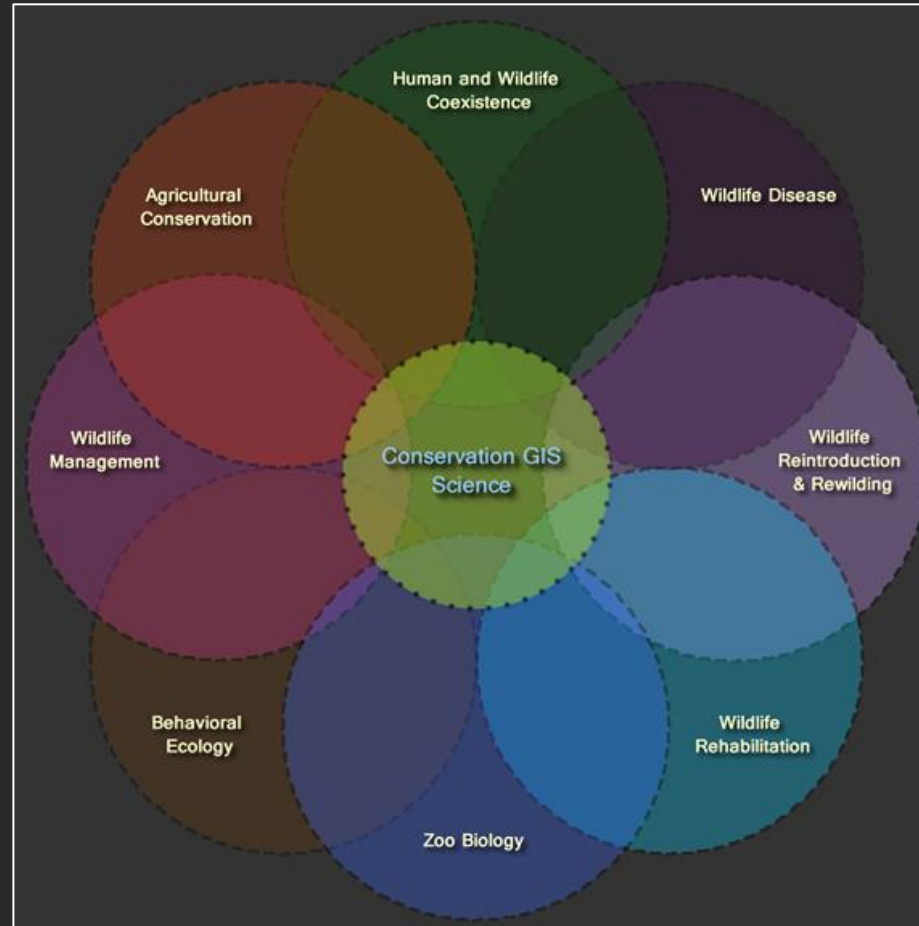
Advisor: Dr. Jitendra Bal (JB) Sharma

Co-Authors: Jill D. Pruetz and Landing Badji



About Me

- I work full-time at Iowa State University in Ames, Iowa, in the Department of Agricultural and Biosystems Engineering as a GIS Analyst for the **Daily Erosion Project** (DEP) and **Agricultural Conservation Planning Framework** (ACPF).
- Agricultural Conservation for improving water quality.
- Master of GIS
- B.S. in Animal Ecology
- B.S. in Biology
- A.S. in Zoo Animal Technology
- www.byclaudette.com



These are the topics that interest me most.



In another life, I was a 'cats & primates' zoo keeper at The Zoo in Gulf Breeze, Florida.



Presentation Overview

- Goals and Objectives
- Background
- Study Area
- Methodology
- Results
- Lessons Learned
- Questions?



Photo Credit: Jill D. Pruetz



Goals and Objectives

Spatial Problem

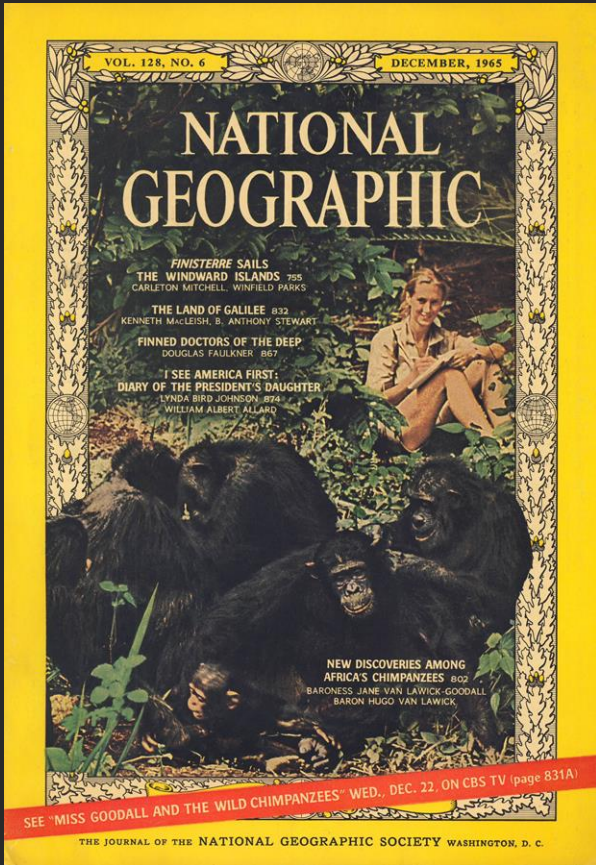
- ALL wild chimpanzees need trees to survive.
- The gallery forests are ecosystems unique to the Sudanian Zone of southeast Senegal that occur along seasonally flooded arterial watercourses.
- These ecosystems provide vital habitat for the critically endangered savanna chimpanzee (*Pan troglodytes verus*) but comprise only 2% of their natural range.
- Today, they are under increasing anthropogenic pressures, primarily from mining, agriculture, and settlement development, and their preservation is becoming critical for savanna chimpanzee conservation.
- This project begins to survey the health of the gallery forests since the expansion of a 10-year gold mining boom in Kedougou, Senegal.
- This study developed a remote sensing method for tracking and mapping gallery forests in Kedougou, Senegal, using Trimble eCognition 10.3 software to apply an unsupervised classification with data fusion and object-based image analysis segmentation.
- This project supports the conservation effort to simultaneously protect gallery forest ecosystems and the critically endangered savanna chimpanzee.





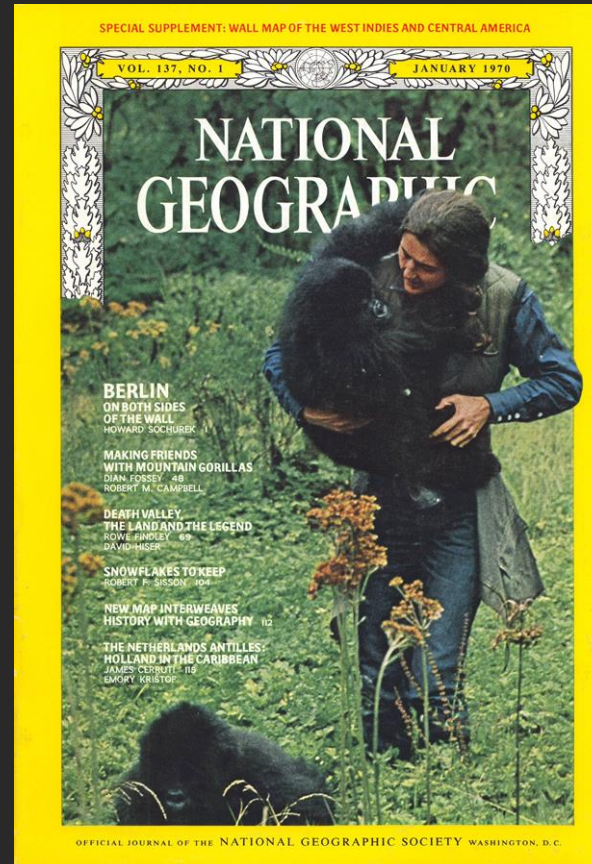
Background

In the beginning, there were the Trimates or Leakey's Angels



JANE GOODALL

Chimpanzees
1960 to present



DIANE FOSSEY

Mountain Gorillas
1967 to 1985



BIRUTE GALDIKAS

Orangutans
1971 to present

Louis Leakey, a famous anthropologist, hired these three women to study great ape behavior in the field.

In many ways, all three were the trailblazers for great ape conservation.

They still inspire people today.

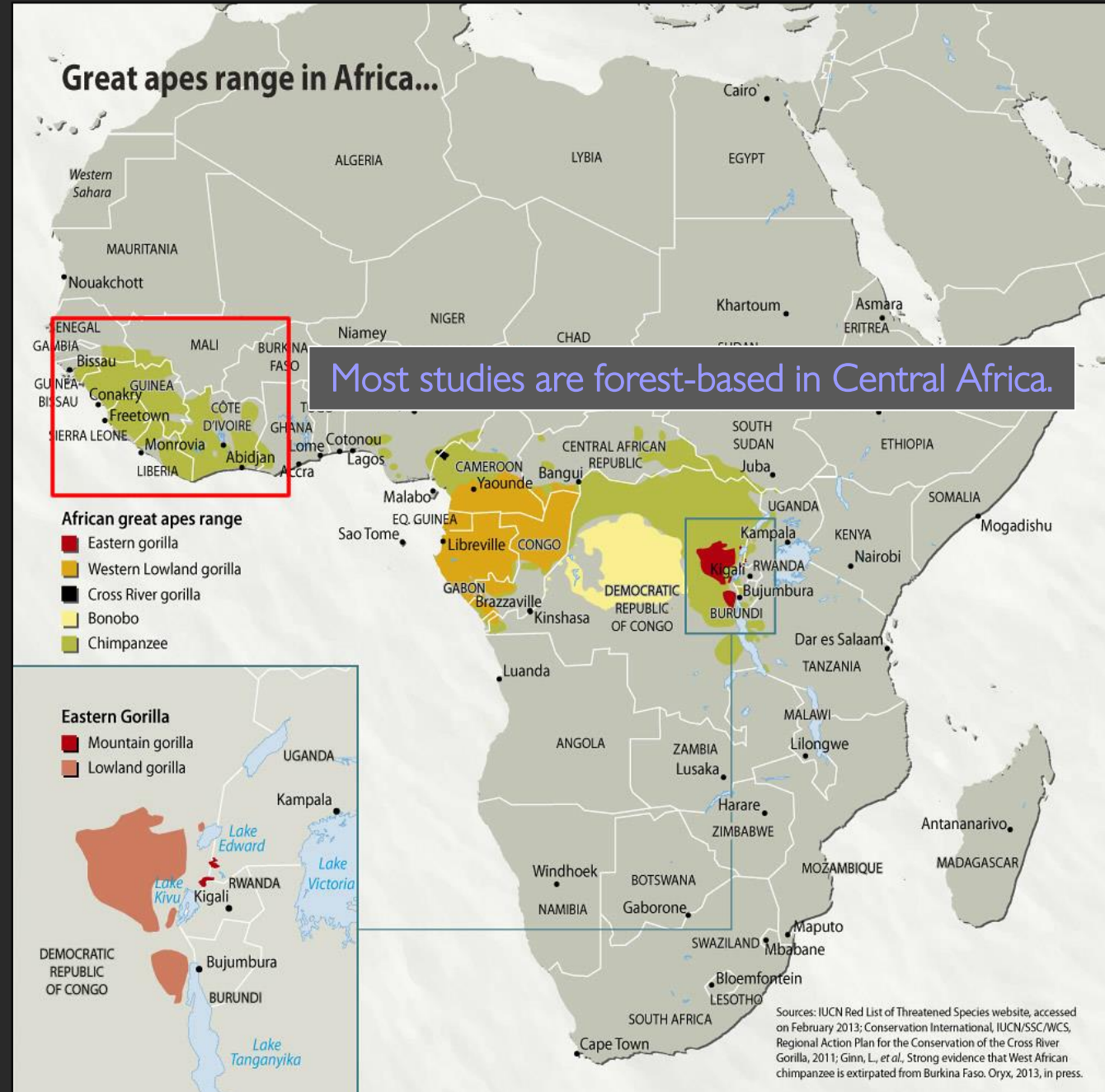


Background

Great Ape Distribution In Africa

There are four officially recognized chimpanzee subspecies:

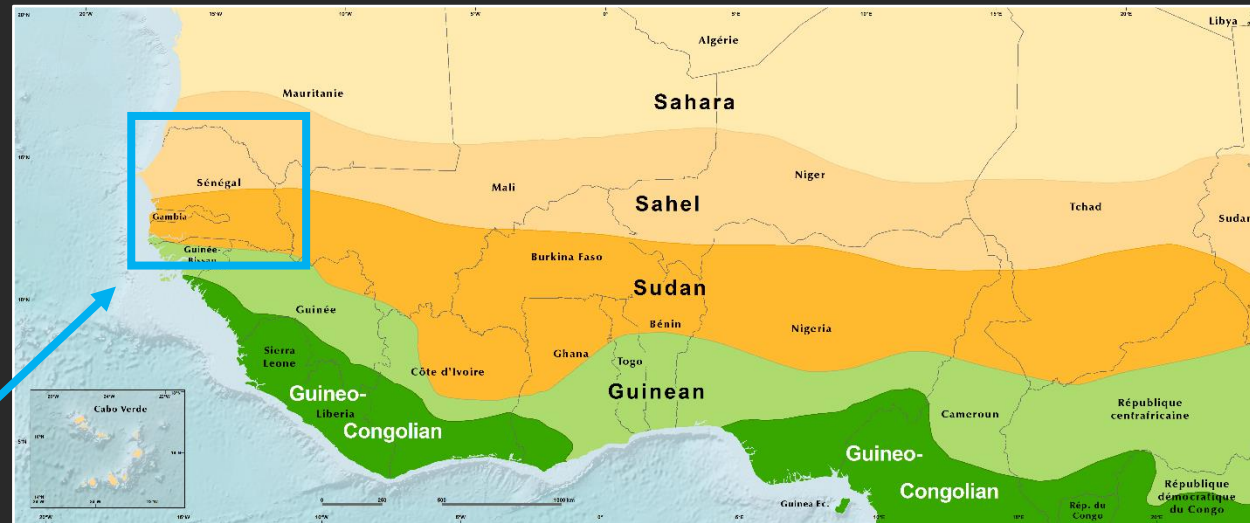
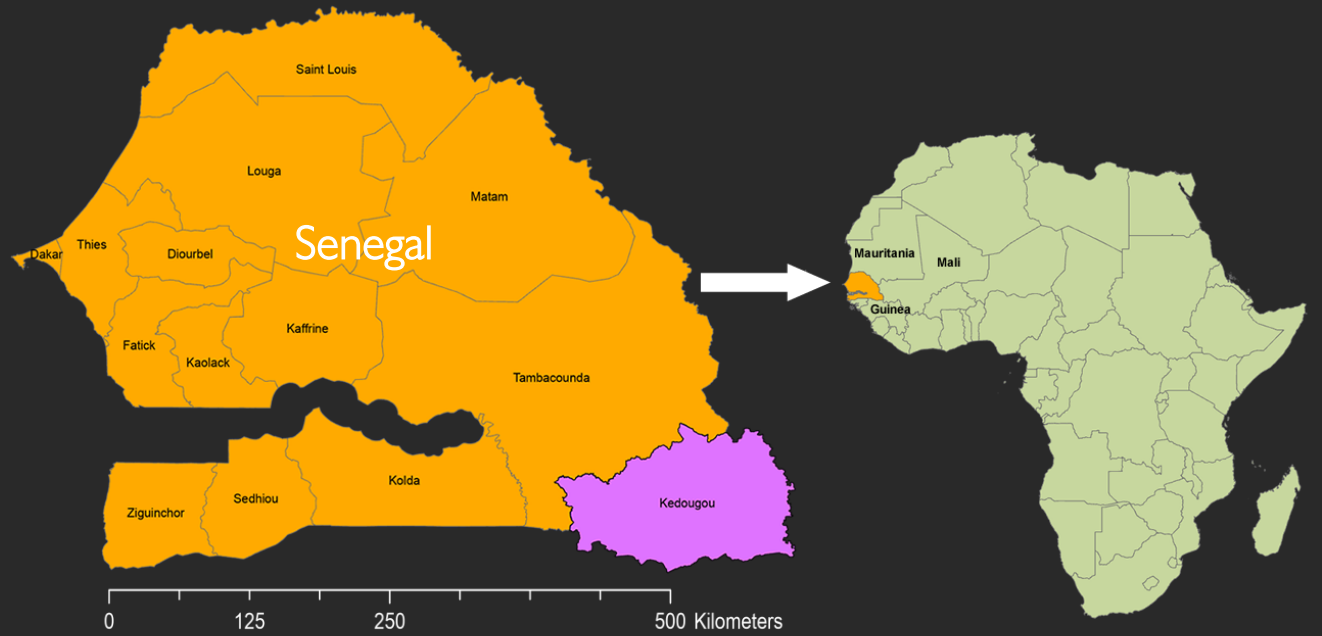
- **Western chimpanzee or Savanna Chimpanzee** (*Pan troglodytes verus*) in the **red box** covering eight West African countries.
- Central chimpanzee (*Pan troglodytes troglodytes*)
- Nigeria-Cameroon chimpanzee (*Pan troglodytes ellioti*)
- Eastern chimpanzee (*Pan troglodytes schweinfurthii*)



Background

Senegal

- Senegal is slightly larger than South Dakota!
- The population is 17.32 million people (2022).
- It is the westernmost country on the African continent.
- Dakar is the capital, and its ports are known as the Gateway to West Africa.
- The region of Kedougou is in the Sudanian Zone and it is characterized as **the domain of the savannas**.



The Bioclimatic Zones of West Africa, the Sahara, Sahel, Sudan, Guinean, and Guineo-Congolian (CILSS, USGS, & USAID, 2016).

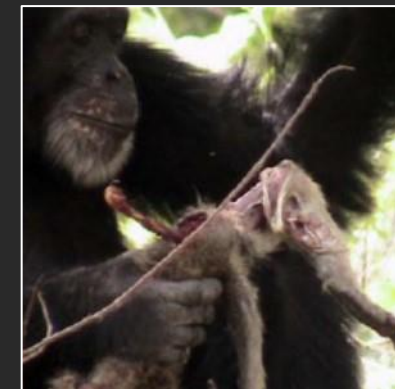
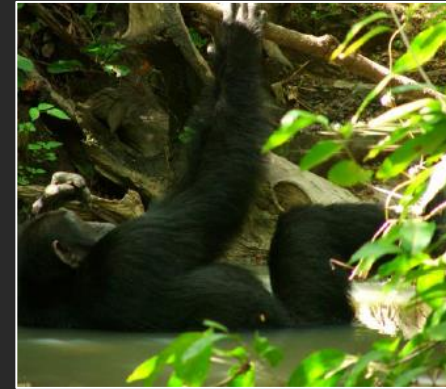




Background

Savanna Chimpanzee (*Pan troglodytes verus*)

- They were listed as critically endangered in 2016 by the IUCN Red List. The next categories are **Extinct in the Wild and Extinct**. The population est. is 18,000 to 65,000, an 80% decline since the 90s.
- Senegal is the northernmost limit and the harshest environment for chimpanzees. The dry season temperatures can reach 40 °C (104 °F).
- They have a unique suite of behaviors adapted for an open, hot, dry, and mosaic environment.
- They termite fish, and hunt galagoes with a spear that they fashion themselves!
- They utilize microclimates such as gallery forests, caves, and pools to cool off during the dry season.
- Only 2% of their habitat is forested.
- They would rather be in a gallery forest during the hottest time of the dry season because this is where water, food, shade, and tall evergreen trees are for nesting and protection from predators.
- Also, the savanna chimpanzee adaptations help us to understand how our last common ancestor survived in an open, hot, dry and mosaic environment.



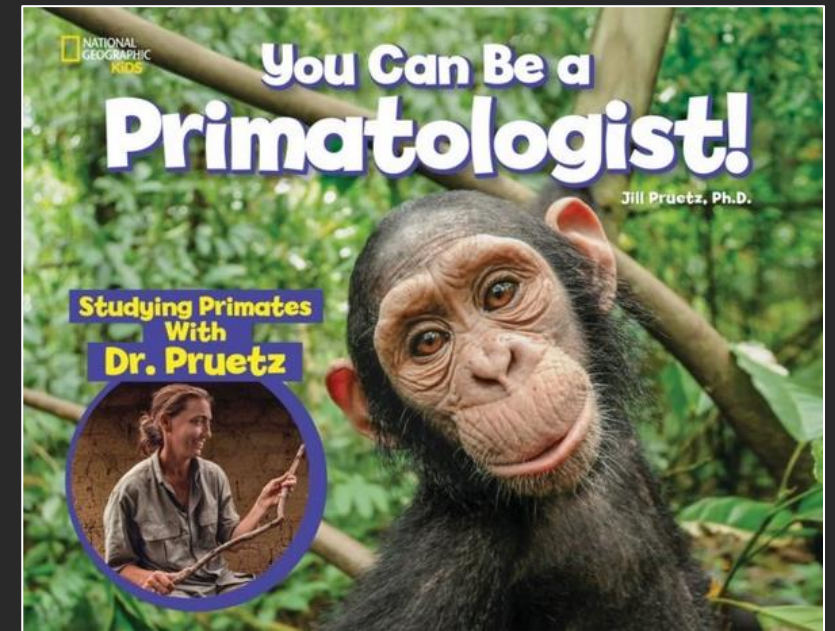


Background

The Famed Fongoli Chimpanzees



As seen on BBC Earth and Max. <https://youtu.be/SN7CLg2LixY>



Dr. Jill Pruetz from Iowa State University to Texas State University is a biological anthropologist who studies savanna chimpanzees at the Fongoli Savanna Chimpanzee Project and the founder.

This project used her nesting data, topography maps, and knowledge.



Background

Gold Mining in Kedougou, Senegal

- There has been a gold mining boom in Senegal for over 14 years.
- The gold mining boom presents a great challenge for primate conservation because the mining brings new levels of anthropogenic disturbances and ecological pressures.
- The disturbances can include loss of group connectivity and loss of connectivity to habitat preference and protected areas — due to road construction, mining pits, pond tailings, fencing, settlement development, and forest degradation.



Gold Price Tracker: <https://goldbroker.com/widget/historical/XAU?height=500¤cy=USD>

Background

Types of Mining

Large-scale Gold Mine

- Industrial mining with trained employees, using large-scale mechanized tools to extract the gold quickly.
- Investors are foreign and West African.



Large-scale: **Mako gold mine**, photo credit: Resolute, reproduced for educational purposes only.

Intermediate-scale Artisanal Gold Mine

- Large-scale Artisanal Mine that has more infrastructure.

Small-scale Artisanal Gold Mines (Djouras)

- Is cultural subsistence mining.
- This is gold panning with iron tools or small power machines.
- Investors are traditional local people.
- Djouras have been culturally present for millennia with local rules.



Intermediate-scale Artisanal Mine: **Bantakocouta gold mine**, photo credit: niokolo.com, reproduced for educational purposes only.



Small-scale Artisanal Mine: Open pit gold mine, photo credit: rivergambiaexpedition, reproduced for educational purposes only.

Background

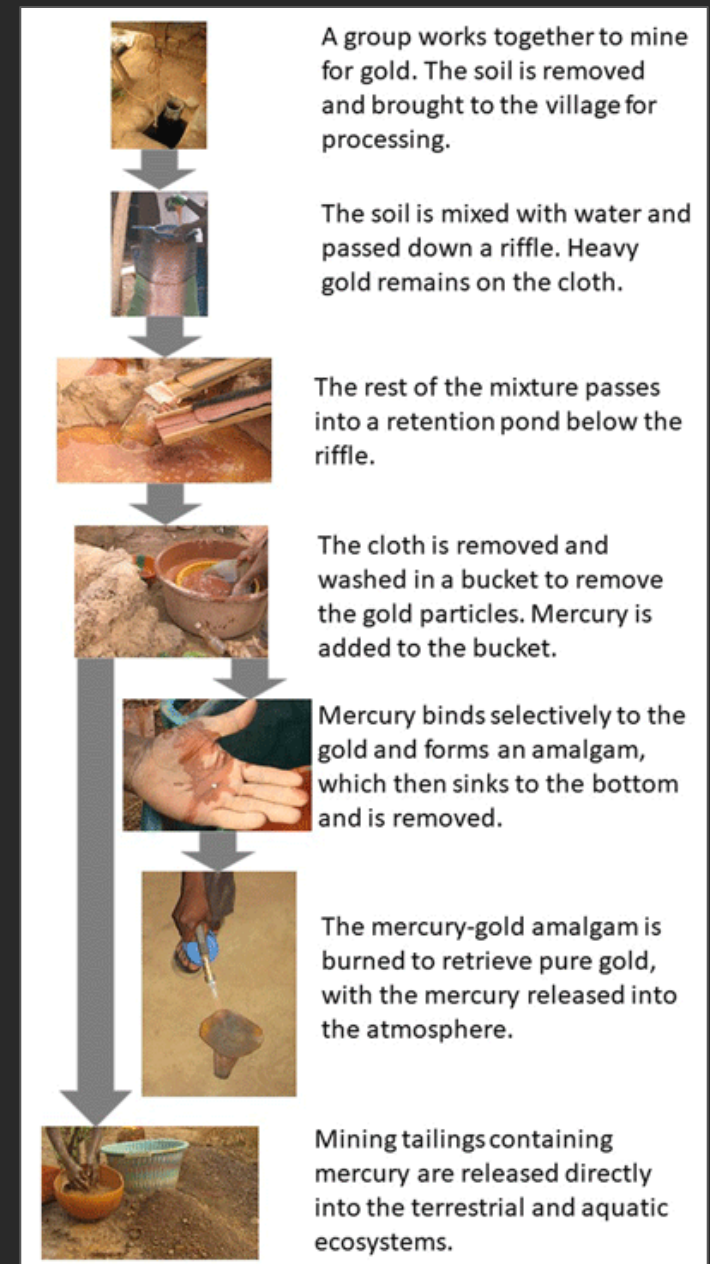
Small-scale Process

Furthermore, gold mining brings mercury-contaminated water to terrestrial and aquatic ecosystems.



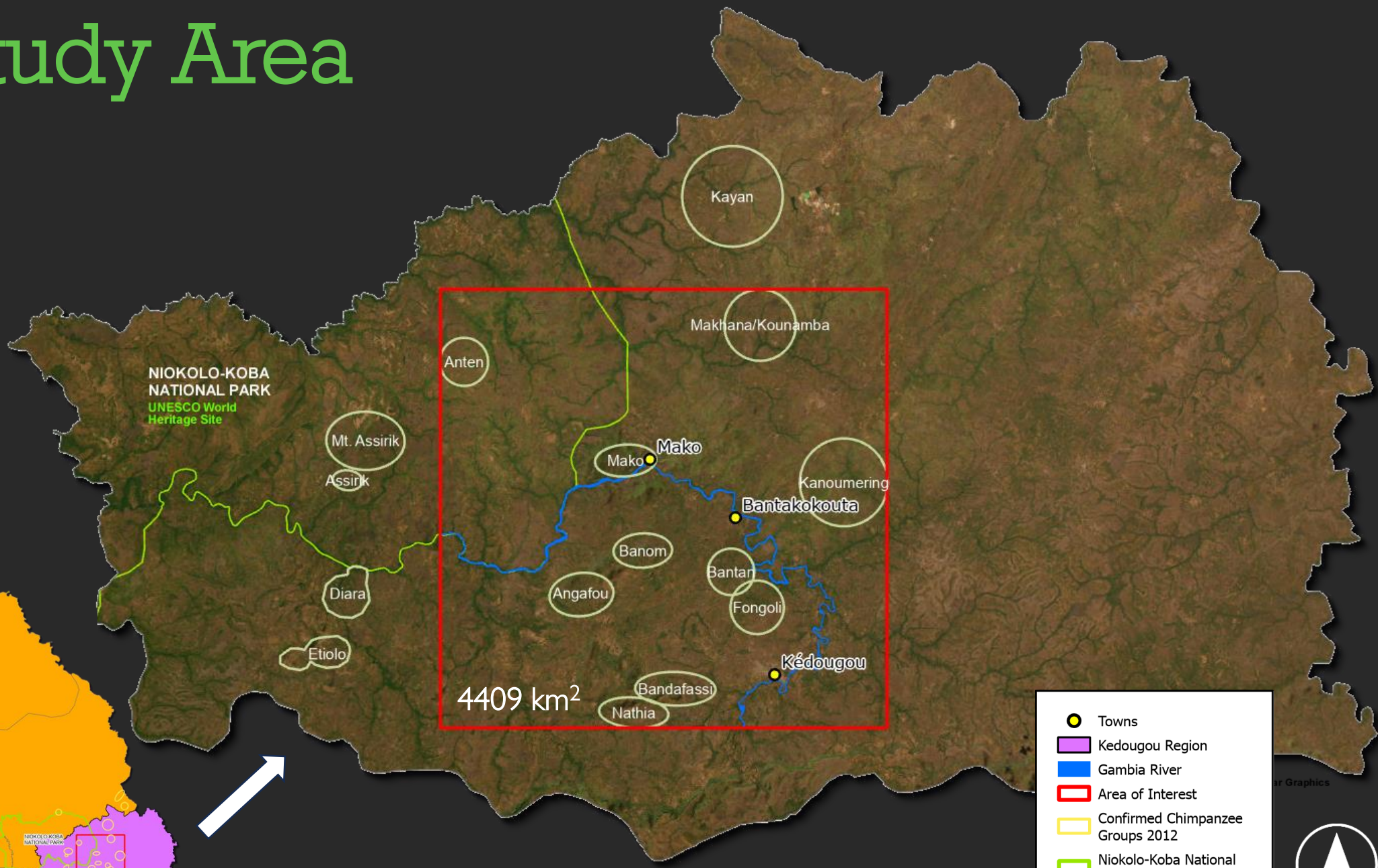
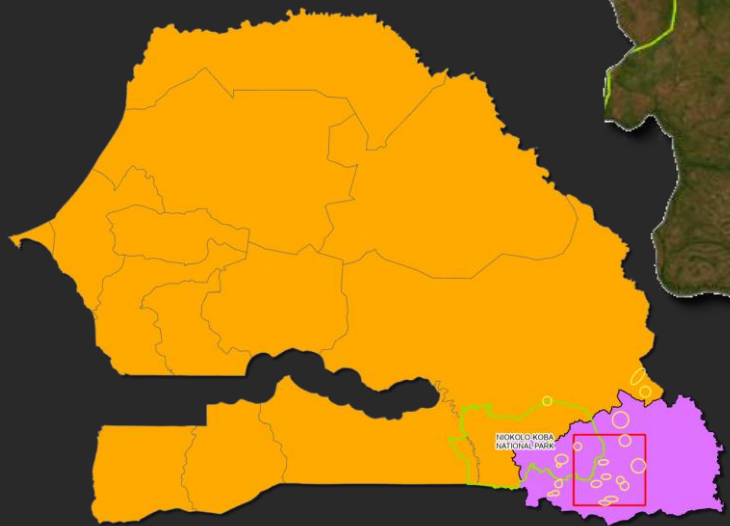
Photo Credit: Jill D. Pruetz

Fongoli savanna chimpanzees drink water from a polluted artisanal mining pit.



The Small-scale Artisanal Gold Mining Process in Kedougou, Senegal (Gerson, Driscoll, Hsu-Kim, & Bernhardt, 2018).

Study Area



- Towns
- Kedougou Region
- Gambia River
- Area of Interest
- Confirmed Chimpanzee Groups 2012
- Niokolo-Koba National Park

ar Graphics





Methodology

Workflow

STAGE 1

Data Acquisition

- SRTM DEM (NASA EarthData), PlanetScope (Planet NICFI Program), Sentinel 2A (ESA Copernicus), Landsat 8, 7, 5 (EarthExplorer).

Pre-processing

- Composite Landsat and Sentinel 2A.
- Mosaic Sentinel 2A, PlanetScope, and SRTM DEM.
- Subset all to the Area of Interest (AOI) and adjust the spatial reference to WGS 1984 UTM Zone 28N.
- Data Exploration

STAGE 2

Build Layers

- Create stream network vector and raster and local and subregional watershed boundaries using ArcGIS Pro and Arc Hydro Toolbox.
- Create soils vector, raster, tables and maps from World Soil Information.
- Create Principal Component Analysis for the AOI in 2023, 2010, 2000, and 1988.
- Create indices for NDWI, NDVI, and SAVI for the AOI for all 4 years in eCognition and export out to save and import back into eCognition.

Build Remotely Sensed Meteorological Data

- Create CHIRPS monthly mean rainfall estimates from 1988 to 2023 specifically for the AOI using Google Earth Engine.
- Create MODIS Land Surface Temperature estimates from 2000 to 2023 specifically for the AOI using Google Earth Engine.

STAGE 3

Classification

- Using Trimble eCognition to classify the AOI for 2023, 2010, 2000, and 1988.
- Ruleset refinement for unsupervised classification and multi-threshold segmentation.
- Classification for: Closed-vegetation, Ecotone, Open-vegetation, Bare Soil, Vegetation Degradation, Development, Large-scale Mine, Mine Tailing Pond, Intermediate-scale Artisanal Mine, Small-scale Artisanal Mine, Water Body, Roads, and NoData.
- Export eCognition vector output to ArcGIS Pro to process and create tables, charts, and maps for the AOI, subregional watershed, and local watershed.

STAGE 4

Accuracy Assessment

- For 2023 classification using ArcGIS Pro and PlanetScope as the reference data.

Change Detection

- Using ArcGIS Pro Change Detection Wizard for 1988 to 2023.

Presence-only Prediction

- Using ArcGIS Pro Presence-only Prediction Tool to predict where the savanna chimpanzee habitat was suitable in 2023.
- Based on savanna chimpanzee nesting data, the classified 2023 AOI, stream network, dominate soils, parent soils, land forms, elevation, aspect, and OpenStreetMap roads.

Methodology

Data

January	February	March	April	May	June	July	August	September	October	November	December
Dry	Dry	Dry	Dry	Transitional	Wet	Wet	Wet	Wet	Transitional	Dry	Dry

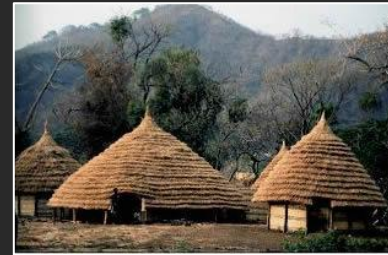


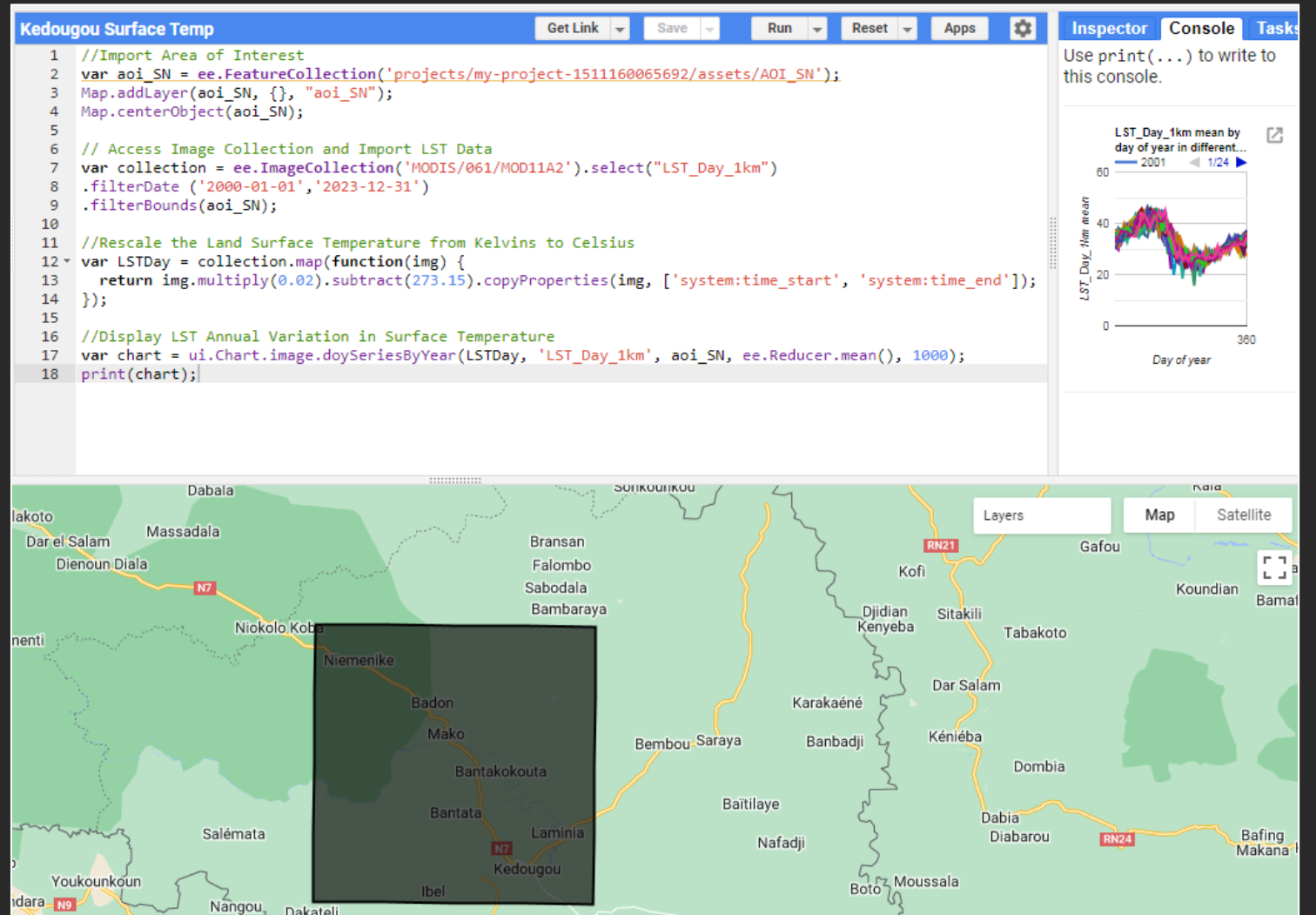
Photo Credit: Gray Tappan

- Senegal's landscape is a highly mosaic environment. Which can make it very difficult to classify the land cover and land use.
- Dry season is from **November to April** and **May** is a transitional month.
- Wet season (rain and clouds) is from **June to September** and **October** is a transitional month.
- Senegal's seasons and vegetation growth are highly correlated with annual precipitation.
- Satellite imagery was selected for the month of **December**.
- December was chosen to get a snapshot of the gallery forests when the highest and lowest growth does not interfere with capturing the classification ... in theory.

Methodology

Google Earth Engine Data

- I used Google Earth Engine (GEE) to analyze land surface temperature and precipitation, which allowed me to observe seasonality over time. This analysis also guided my decision to select December for my imagery.
- Accessing large datasets on GEE can be accomplished with some simple code.
- Here is an example of how to retrieve land surface temperature data using MODIS.
- I did not write this JavaScript code myself; instead, I followed a video tutorial and made some adjustments.





Methodology

Google Earth Engine Data

- Here is an example of the code for CHIRPS Precipitation.
- You can use the QR Code below to access this information.



Precipitation_1988_2023
Get Link Save Run Reset Apps

```

1 //Extract monthly rainfall from CHIRPS
2
3
4 var aoi_SN = ee.FeatureCollection('projects/my-project-1511160065692/assets/AOI_SN');
5 Map.addLayer(aoi_SN, {}, "aoi_SN");
6 Map.centerObject(aoi_SN);
7
8 //List of years and months
9
10 var years = ee.List.sequence(1988, 2023);
11 var months = ee.List.sequence(1, 12);
12
13 var rainfall = ee.ImageCollection("UCSB-CHG/CHIRPS/DAILY").select('precipitation');
14 print(rainfall.first());
15
16 ///Map over the years and create a monthly totals collection
17 var monthlyImages = years.map(function(year) {
18   return months.map(function(month) {
19     var filtered = rainfall
20       .filter(ee.Filter.calendarRange(year, year, 'year'))
21       .filter(ee.Filter.calendarRange(month, month, 'month'));
22     var monthly = filtered.sum();
23     return monthly.set({'month' : month, 'year' : year});
24   });
25 }).flatten()
26
27 var monthlyCol = ee.ImageCollection.fromImages(monthlyImages);
28 print(monthlyCol);
29
30 //Zonal Statistics to Summarize Rainfall
31 var rainfallKedougou = monthlyCol.map(function(img) {
32   var features = aoi_SN.map(function(f) {return f.set('month', img.get('month'), 'year', img.get('year'))});
33   var proj = ee.Image(monthlyCol.first()).projection();
34   return img.reduceRegions(features, ee.Reducer.mean(), 1000, proj);
35

```

Inspector Console Task
Layers Map Satellite

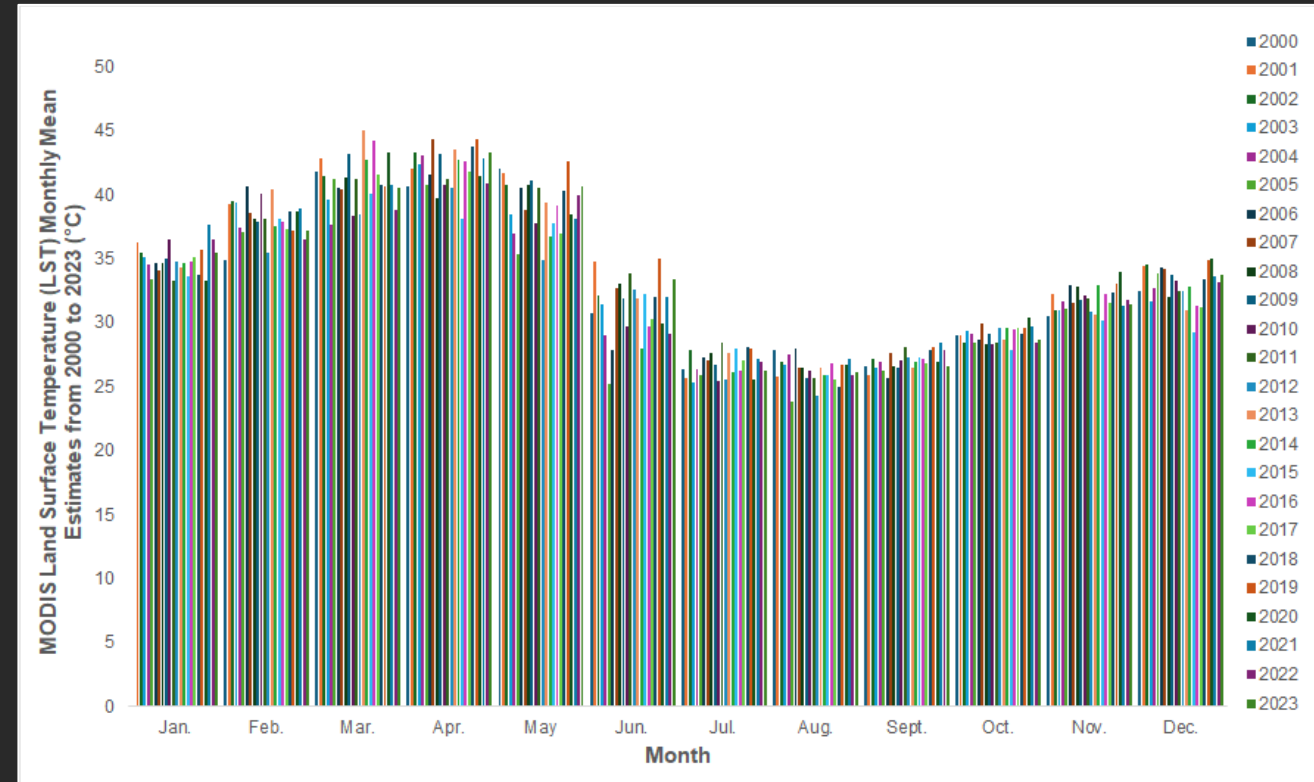
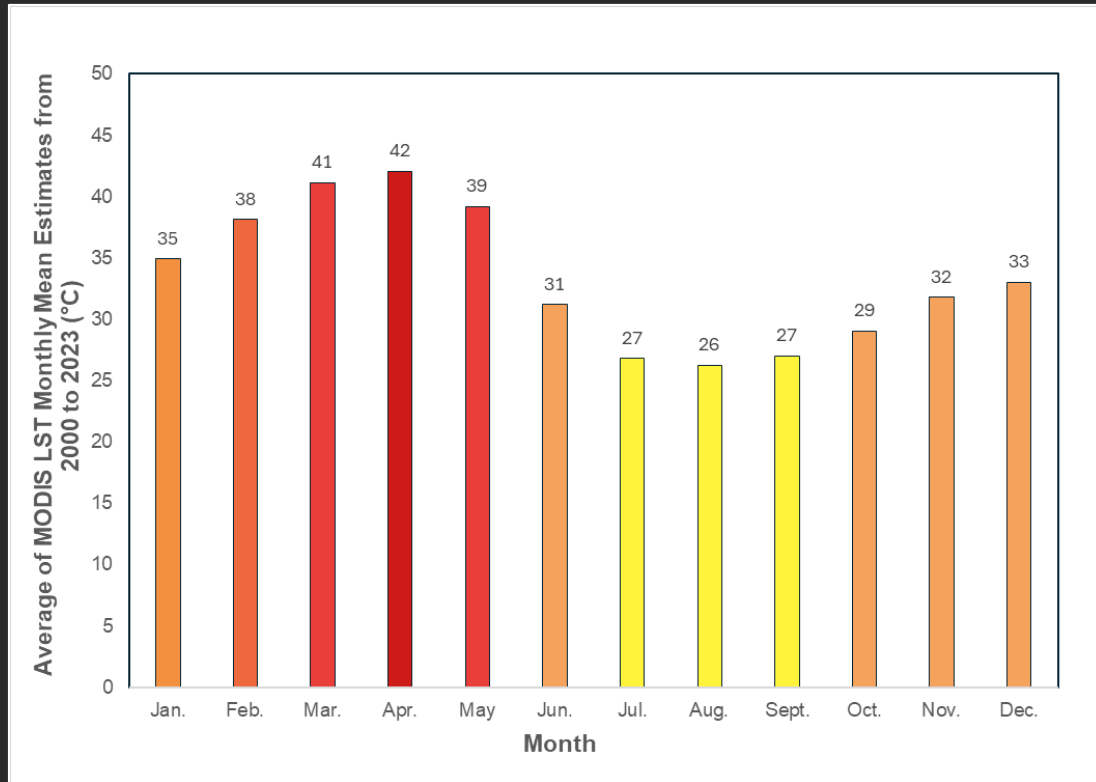
Use print(...) to write to this console.

- ▶ Image UCSB-CHG/C... JSON
- ▶ ImageCollection ... JSON
- Rainfall Summary ... JSON
- ▶ FeatureCollectio... JSON



Methodology

MODIS Land Surface Temperature Data from 2000 to 2023



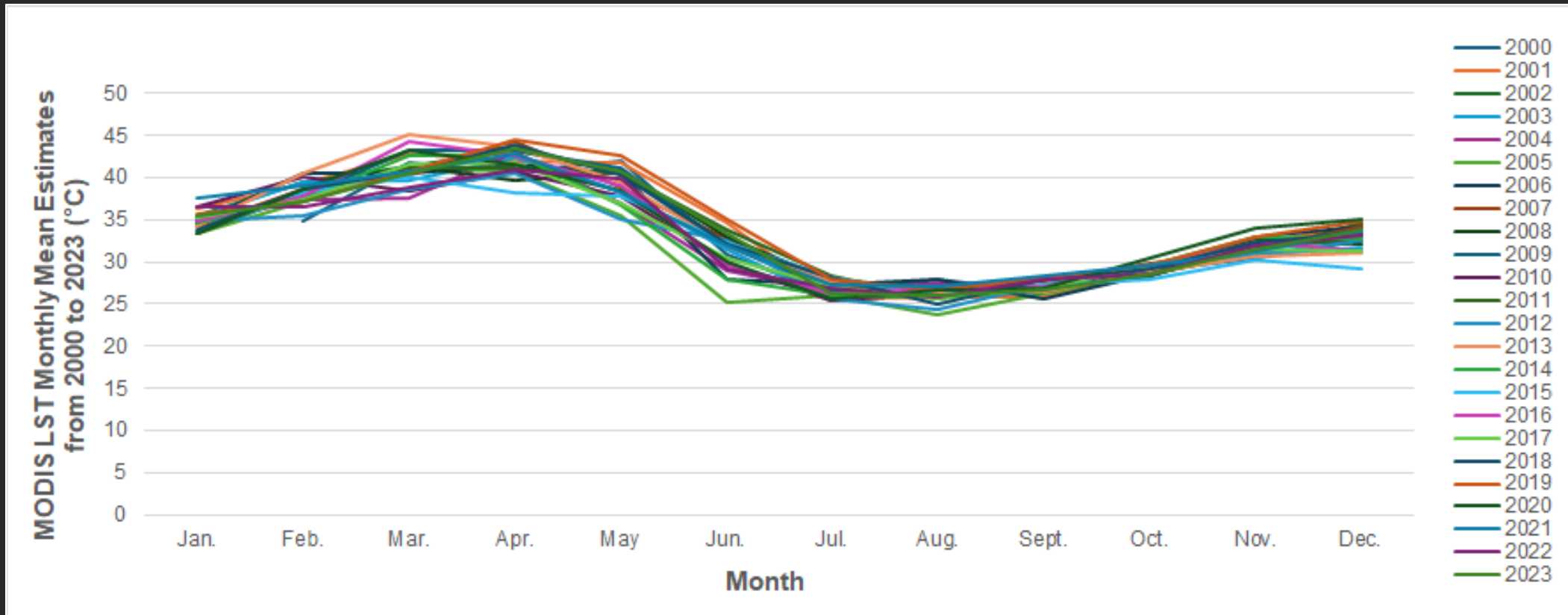
It is hot in Senegal! Hot in the dry season and then it cools down a bit in the wet.

MODIS LST takes an 8-day average of the surface temperature every 8 days.



Methodology

MODIS Land Surface Temperature Data from 2000 to 2023

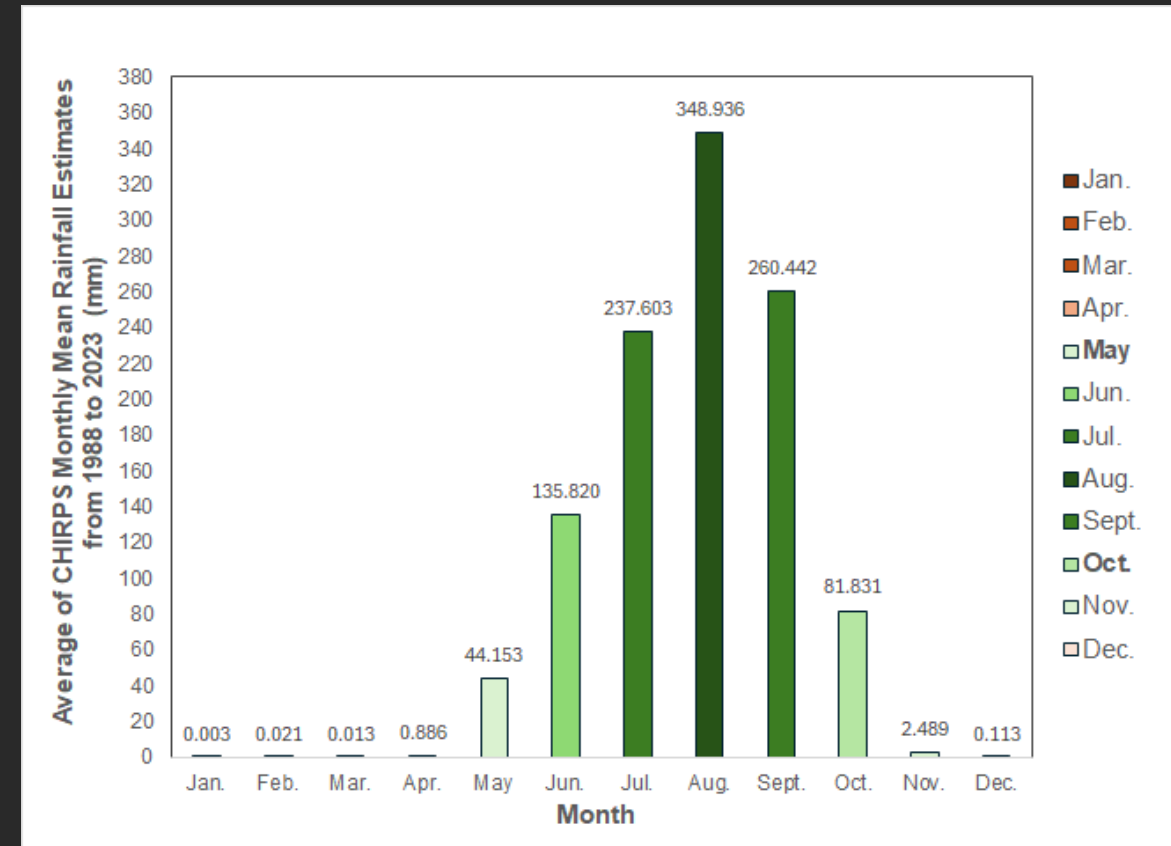
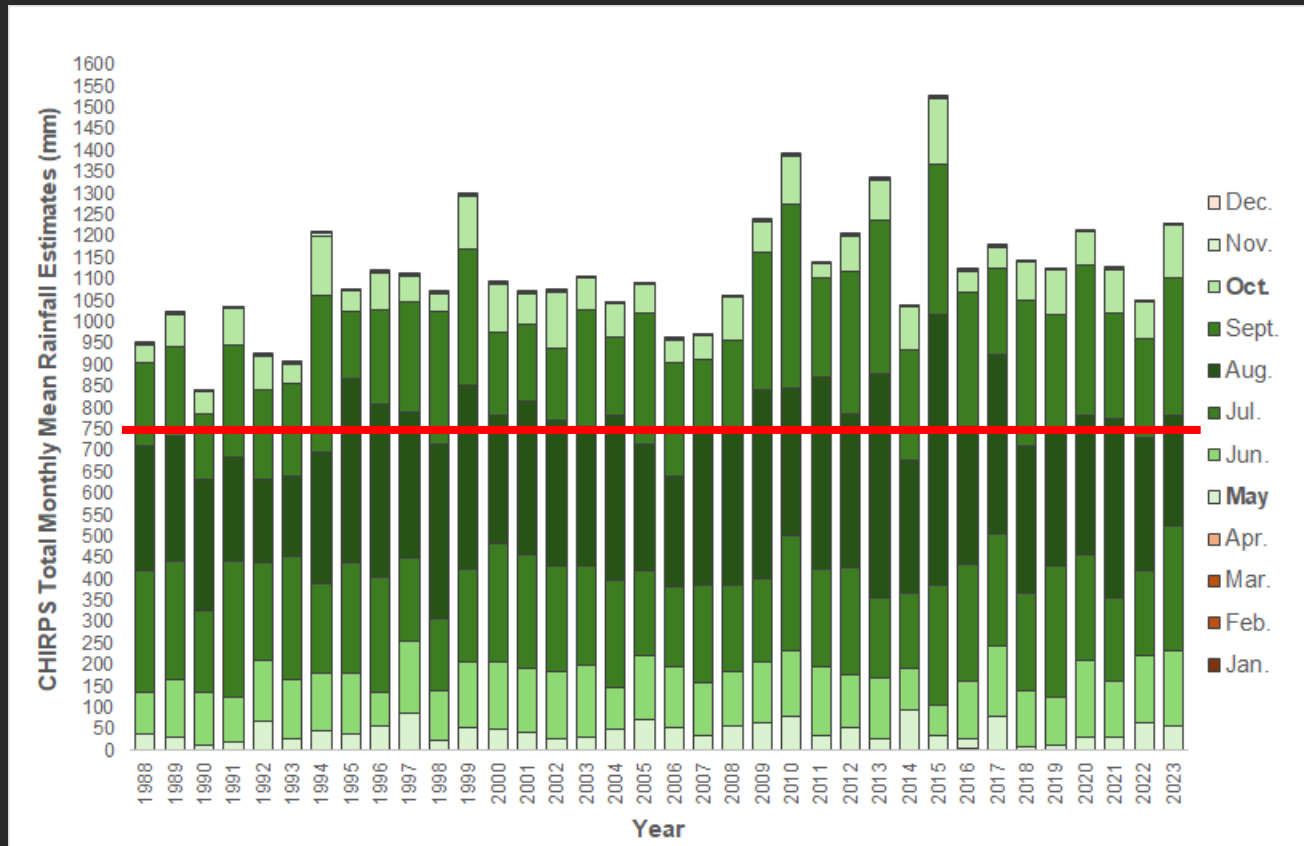


Here is another interpretation that is essentially the same seasonal curve.



Methodology

CHIRPS Precipitation Data for 1988 to 2023



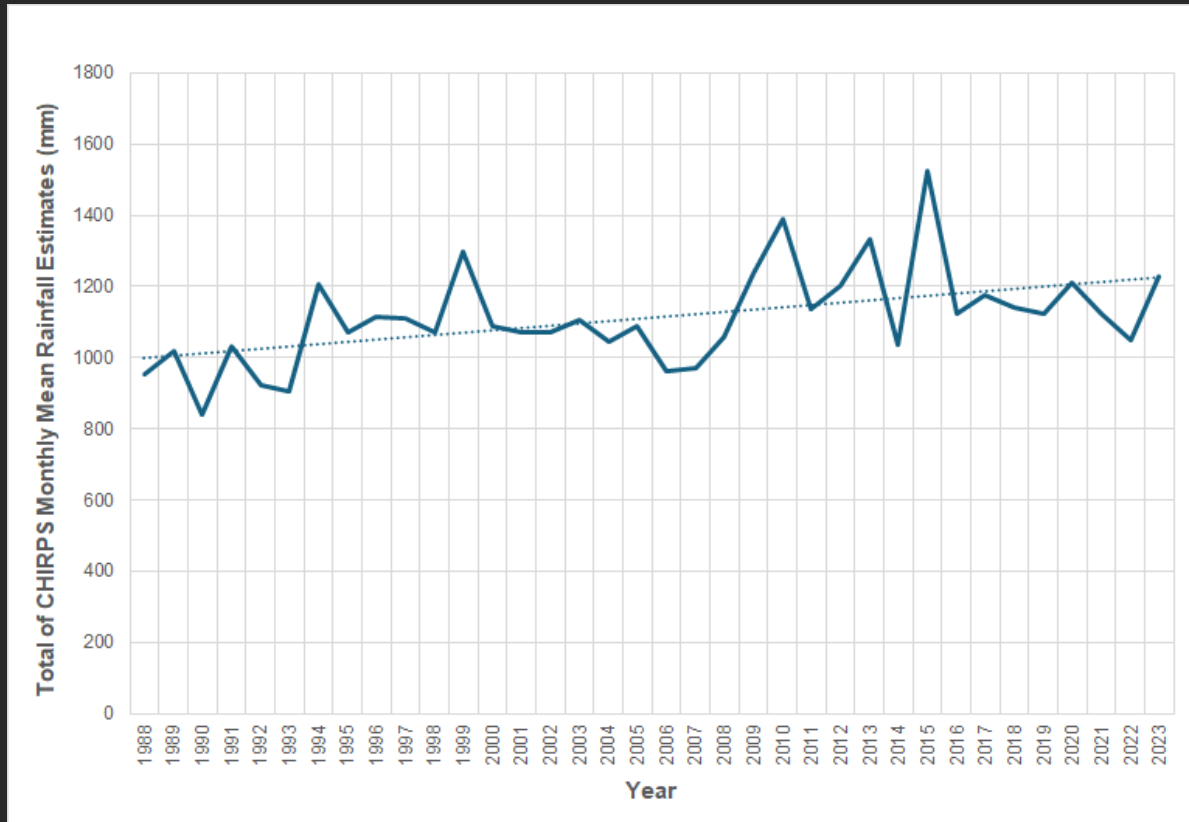
As a reference, chimpanzees have not been observed in areas where the average annual rainfall is less than **750 mm** (Lindsheild, et al., 2021).

The rainfall and vegetation growth are closely correlated so you can infer some seasonality with precipitation.

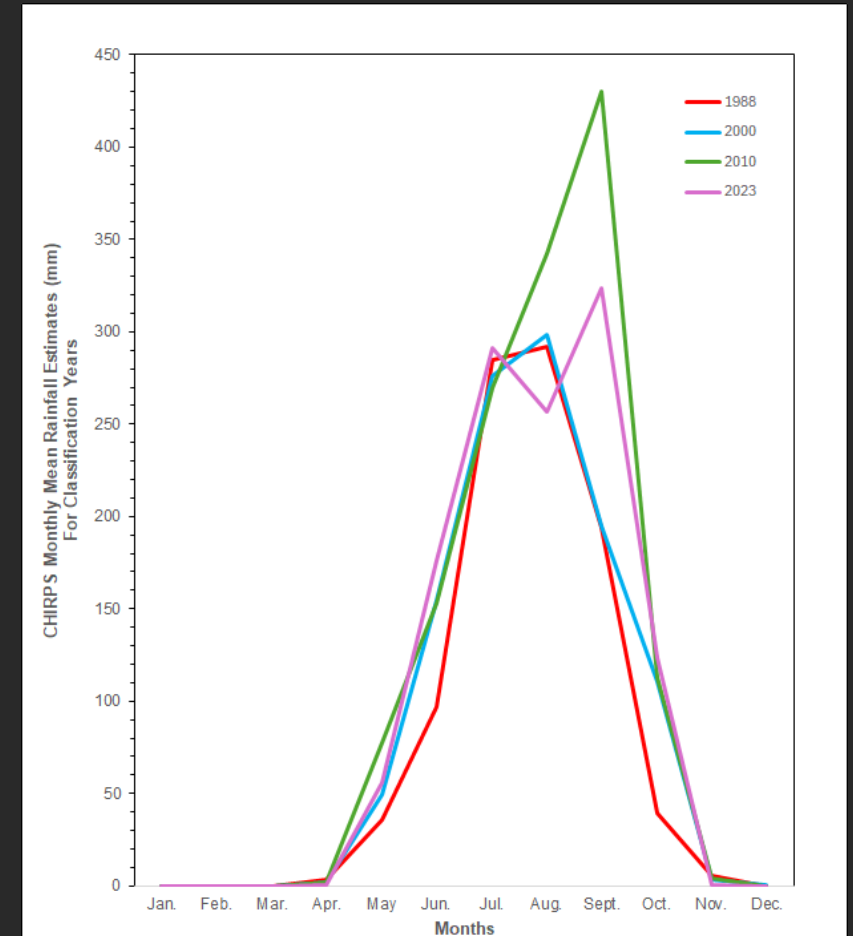


Methodology

CHIRPS Precipitation Data for 1988 to 2023



It looks like they are getting more rain perhaps flooding.



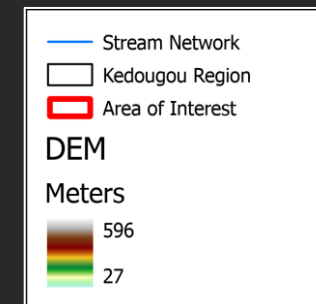
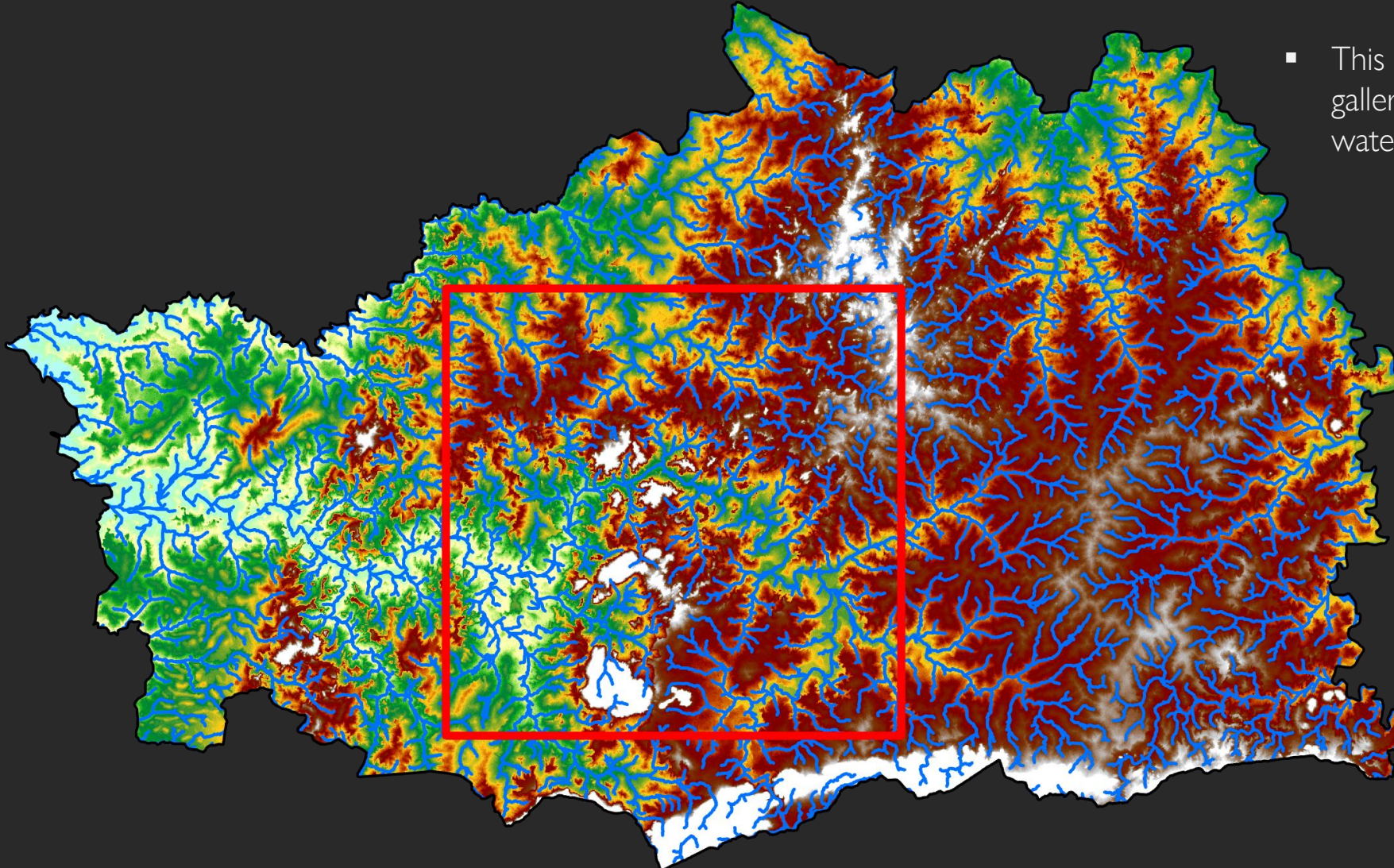
These data are specific to the study area. The 1988, 2000, 2010, and 2023 study areas show three different patterns.



Methodology

Layer Creation: Stream Network

- The stream network was derived from the 30-meter SRTM DEM using ArcGIS Pro and Arc Hydro Toolbox.
- This layer was vital because it helped locate gallery forests predominately found along watercourses.

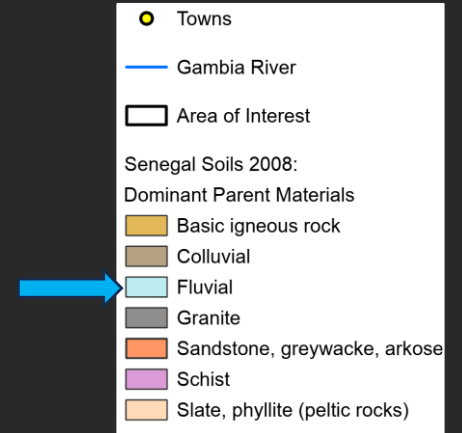
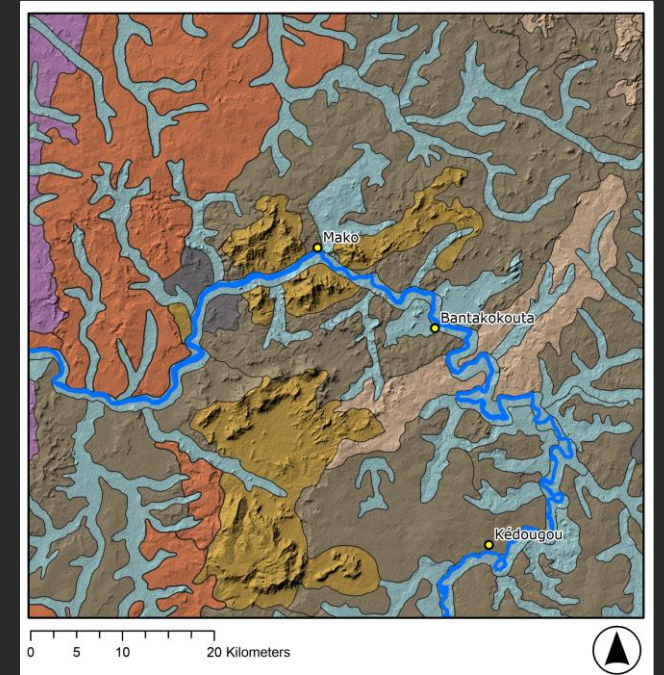
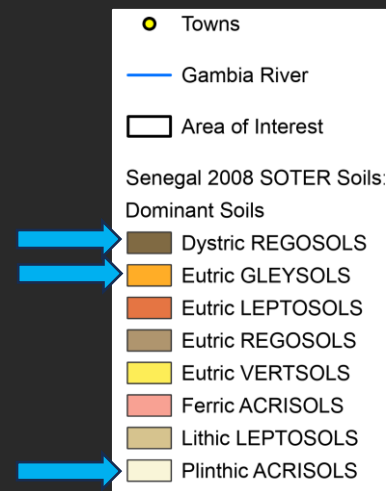
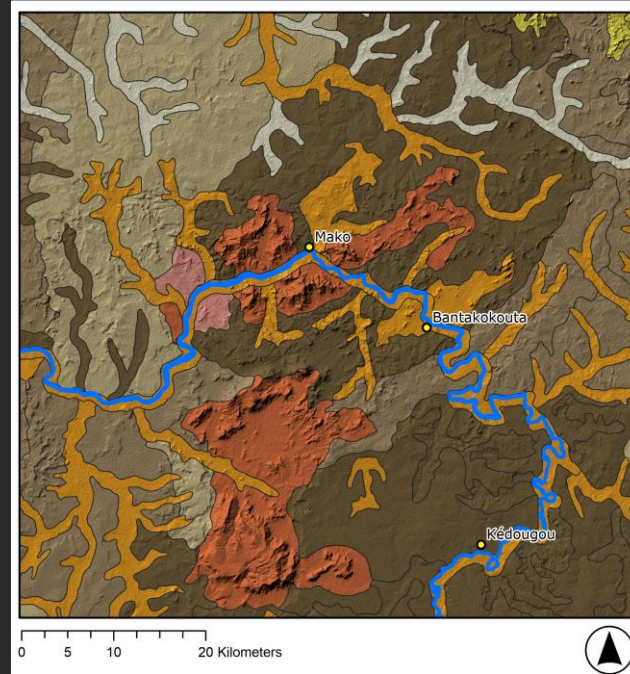
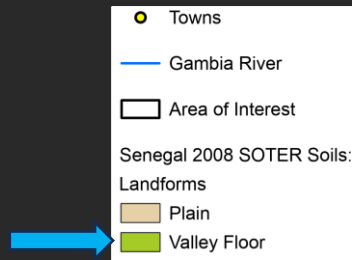
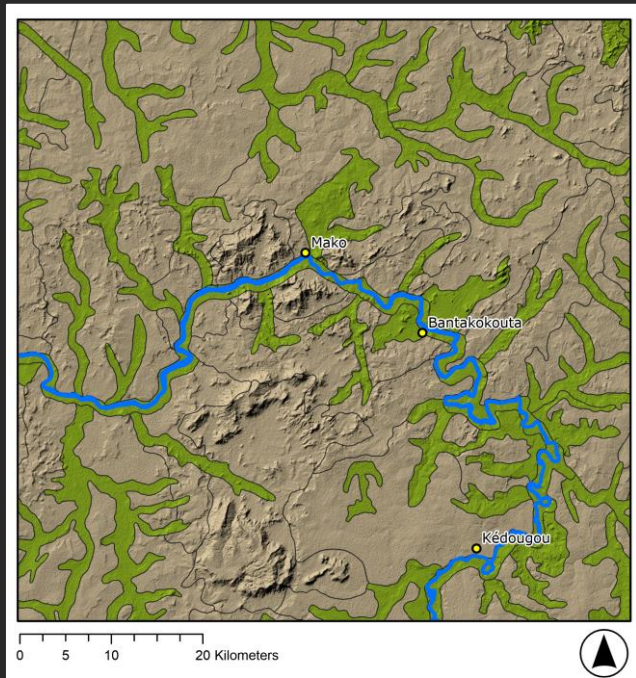




Methodology

Layer Creation: Soils

The soils data was acquired from the Soil and Terrain Database (SOTER) for Senegal and the Gambia.

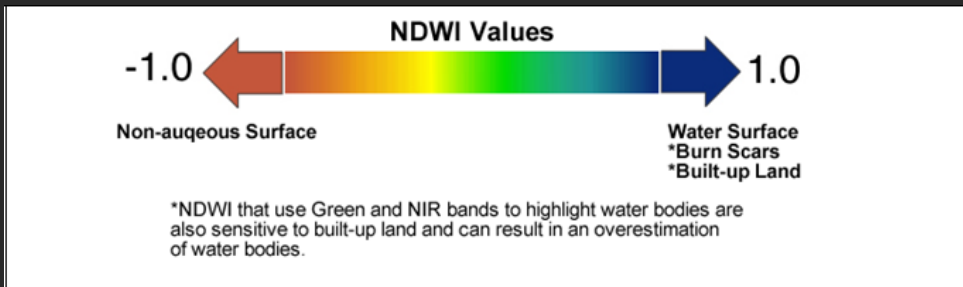
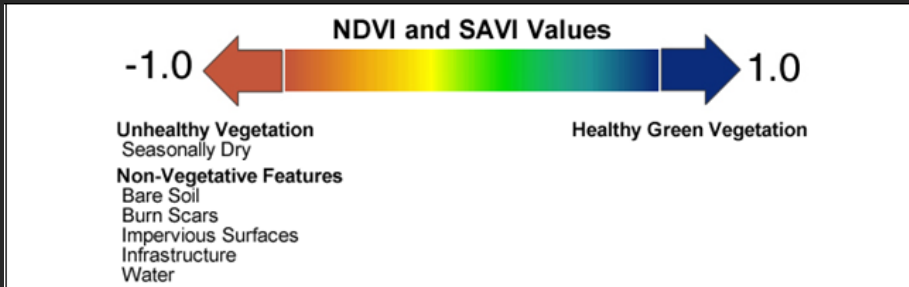




Methodology

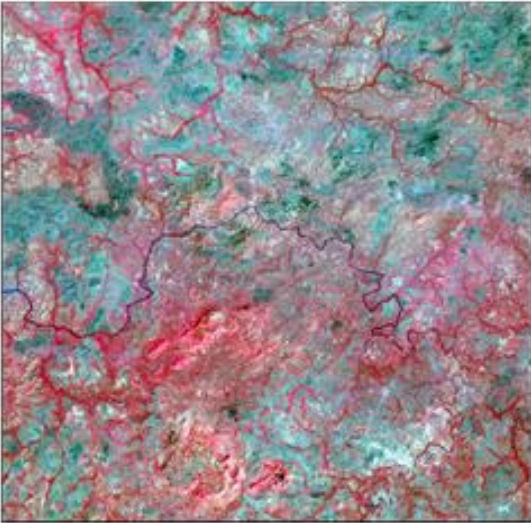
Layer Creation: Indices and PCA

- The PCA colors are random, and feature information is clustered and extruded to show the most dominant features in the scene.
- The indices were built within eCognition by using its canned formulas.

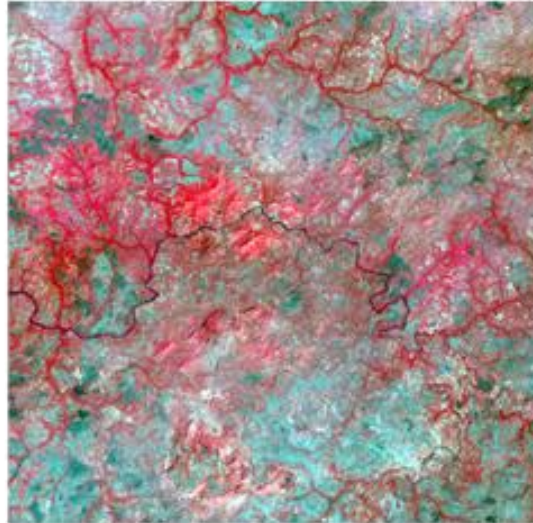


Year	Res.	NDVI (NIR-Red)/(NIR+Red)	SAVI ((NIR-Red)/(NIR+Red+L)) * (1+L) L=0.75	NDWI (Green-NIR)/(Green+NIR)	Principal Component Analysis (PCA)
Dec 2023 Sentinel 2A	10 meters				
Dec 2010 Landsat 5 TM	30 meters				
Dec 2000 Landsat 7 ETM	15 meters				
Dec 1988 Landsat 5 TM	30 meters				

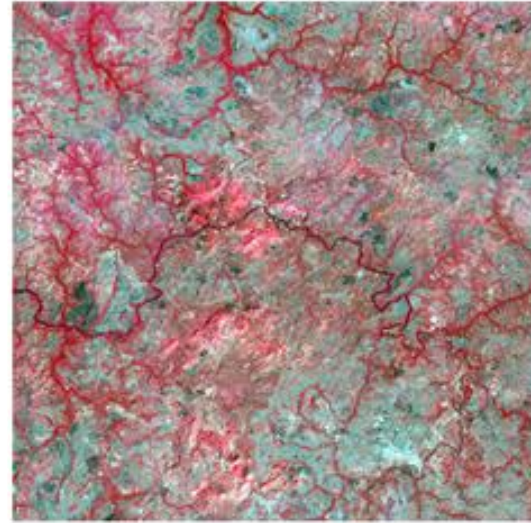
10 December 1988
Landsat 5 TM, C2L2
30 meters



19 December 2000
Landsat 7 ETM, C2L1
15 meters



23 December 2010
Landsat 5 TM, C2L2
30 meters



27 December 2023
Sentinel-2A
10 meters



This is the primary satellite imagery used to set the resolution for each eCognition project. Landsat 7 ETM and Landsat 5 TM used B4, B3, and B2, and Sentinel 2A used B8, B4, and B3 (RGB) to display healthy vegetation in color infrared.



Methodology

eCognition Data Fusion

This is like **supercharging** an unsupervised classification!

Remember, mosaic landscapes are difficult to classify.

An unsupervised classification may be a better choice if you are unfamiliar with the area.

eCognition Data for 1988	eCognition Data for 2000	eCognition Data for 2010	eCognition Data for 2023
<p>Landsat 5 TM, C2L2 Tier 1, 16 Bit, 30 m, Date: 12/10/1988 PCA of Landsat 5</p> <p>DEM, 30 m, Date: 2/11/2000</p> <p>Constructed in eCognition with Landsat SAVI_1988.tif NDVI_1988.tif NDWI_1988.tif</p> <p>Overlays in eCognition for Classifying Soils, Vector, Date: 2008 Stream Network, Vector, Date: 2/11/2000 Artisanal Mines, Vector, 2014 OpenStreetMap Roads Fishnet Guide</p> <p>Reference Data: N/A</p>	<p>Landsat 7 ETM C2L1, 15 m, Date: 12/19/2000 Panchromatic PCA of Landsat 7 C2L2 Pansharpened</p> <p>Landsat 7 ETM C2L2 Tier 1, 16 Bit, 30 m, Date: 12/19/2000 PCA of Landsat 7 C2L2</p> <p>DEM, 30 m, Date: 2/11/2000</p> <p>Constructed in eCognition with Landsat SAVI_2000.tif NDVI_2000.tif NDWI_2000.tif</p> <p>Overlays in eCognition for Classifying Soils, Vector, Date: 2008 Stream Network, Vector, Date: 2/11/2000 Artisanal Mines, Vector, 2014 OpenStreetMap Roads Fishnet Guide</p> <p>Reference Data: N/A</p>	<p>Landsat 5 TM, C2L2 Tier 1, 16 Bit, 30 m, Date: 12/23/2010 PCA of Landsat 5</p> <p>DEM, 30 m, Date: 2/11/2000</p> <p>Constructed in eCognition with Landsat SAVI_2010.tif NDVI_2010.tif NDWI_2010.tif</p> <p>Overlays in eCognition for Classifying Soils, Vector, Date: 2008 Stream Network, Vector, Date: 2/11/2000 Artisanal Mines, Vector, 2014 OpenStreetMap Roads Fishnet Guide</p> <p>Reference Data: N/A</p>	<p>Sentinel2A, 10 m Date: 12/27/23 PCA of Sentinel2A</p> <p>DEM, 30 m, Date: 2/11/2000</p> <p>Constructed in eCognition with Sentinel SAVI_2023.tif NDVI_2023.tif NDWI_2023.tif</p> <p>Overlays in eCognition for Classifying Soils, Vector, Date: 2008 Stream Network, Vector, Date: 2/11/2000 Artisanal Mines, Vector, 2014 OpenStreetMap Roads Fishnet Guide</p> <p>Reference Data: PlanetScope 4.47 m Date: 12/27/23</p>



Methodology

eCognition Classification Ruleset

1. Preprocessing indices.
2. Run **unsupervised classification**. This can take 20 to 30 minutes.
3. Run Create /Update Generic Classes.
4. Run **Multi-threshold Segmentation** with the layer produced from the unsupervised classification. This creates the number of clusters (or classes) you assign.
5. Manually assign classes.
6. Refine the classification.
7. Export.

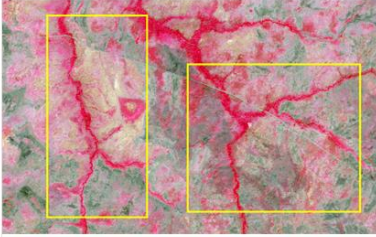
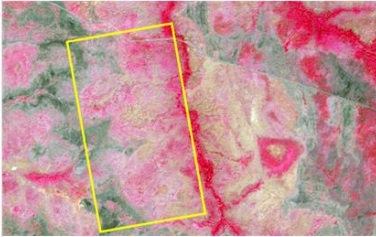
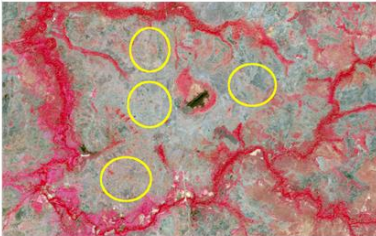
Process Tree


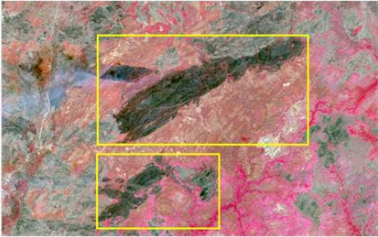

- Ruleset
 - PREPROCESSING
 - index layer NDVI 'NDVI_2023' (S2A_Red, S2A_NIR)
 - index layer NDWI 'NDWI_2023' (S2A_Green, S2A_NIR)
 - index layer SAVI 'SAVI_2023' (S2A_NIR, S2A_Red)
 - UNSUPERVISED CLASSIFICATION
 - unsupervised classification (ISODATA): [DEM_SRTM, S2A_Blue, S2A_Green, S2A_NIR, S2A_PCA_1, S2A_PCA_2, S2A_PCA_3, S2A_PCA_4, S2A_Red] -> SN_2023e (num iterations=35,max clusters=15, min cluster size=50)
 - export image to to ExportImage
 - CREATE OR UPDATE GENERIC CLASSES
 - 15x: create/update class "<auto>"[-,superclass=-,group=-,rgb=-1,-1,-1,scope=Global]
 - MULTI-THRESHOLD SEGMENTATION
 - delete 'New Level'
 - 15x: multi-threshold: creating 'New Level': Class.104 <= 1 < Class.105 <= 2 < Class.106 <= 3 < Class.107 <= 4 < Class.108 <= 5 < Class.109 <= 6 < Class.110 <= 7 < Class.111 <= 8 < Class.112 <= 9 < Class.113 <= 10 < Class.114 <= 11 < Class.115 <= 12 < Class.11
 - ASSIGN CLASSES
 - Class.106 at New Level: Closed-vegetation
 - Class.111 at New Level: Ecotone
 - Class.107 at New Level: Open-vegetation
 - Class.105 at New Level: NoData
 - Class.114 at New Level: Artisanal Small-scale Mine
 - Class.113 at New Level: Bare Soil
 - Class.110 at New Level: Water Bodies
 - Class.109 at New Level: Burn Scars
 - Class.112 at New Level: Development
 - Class.104 at New Level: Other
 - Class.108 at New Level: Roads
 - Class.115 at New Level: Large-scale Mine
 - Class.116 at New Level: Mine Tailing Ponds
 - Class.117, Class.118 at New Level: Intermediate-scale Artisanal Mine
 - REFINE
 - manual classification (brush: 5) Water Bodies -> Burn Scars
 - Artisanal Small-scale Mine, Bare Soil, Burn Scars, Closed-vegetation, Development, Ecotone, Intermediate-scale Artisanal Mine, Large-scale Mine, Mine Tailing Ponds, NoData, Open-vegetation, Roads, Water Bodies at New Level: merge region
 - Closed-vegetation with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Ecotone with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Open-vegetation with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Bare Soil with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Burn Scars with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Development with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Roads with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Water Bodies with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Large-scale Mine with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Mine Tailing Ponds with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Artisanal Small-scale Mine with Area <= 10 Pxl at New Level: remove objects (merge by shape)
 - Artisanal Small-scale Mine, Bare Soil, Burn Scars, Closed-vegetation, Development, Ecotone, Intermediate-scale Artisanal Mine, Large-scale Mine, Mine Tailing Ponds, NoData, Open-vegetation, Roads, Water Bodies at New Level: merge region
 - EXPORT
 - at New Level: export object shapes to SN_2023_Classification



Methodology

eCognition Visual Classification Key

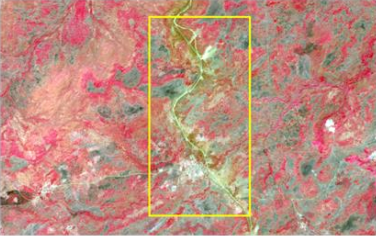
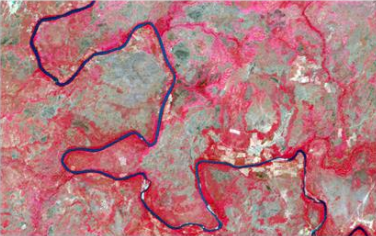
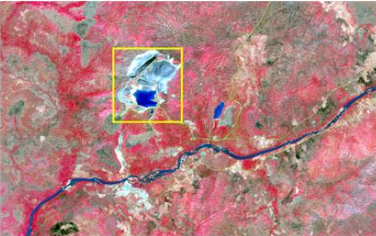
	Classification	Description	Imagery Examples in Infrared	Scale	1988	2000	2010	2023
1	Closed-vegetation (Gallery Forest)	"Closed and evergreen (e.g., gallery/riparian or thicket forest; hereafter "closed vegetation")" (Lindshield, et al., 2021).		1:38,000	✓	✓	✓	✓
2	Ecotone	"A transitional "ecotone" category for vegetation that is neither mostly open nor mostly closed" (Lindshield, et al., 2021).		1:24,000	✓	✓	✓	✓
3	Open-vegetation	"Open and deciduous (e.g., woodland, wooded grassland, and grassland; referred to as "open vegetation")" (Lindshield, et al., 2021).		1:38,000	✓	✓	✓	✓


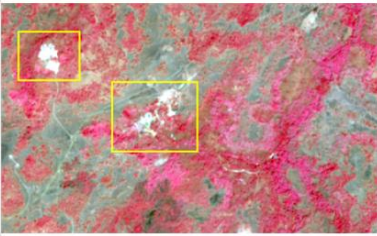

4	Bare Soil	Land with little or no vegetation cover exposing soil, and sandy areas (CILSS, 2016).		1:24,000	✓	✓	✓	✓
5	Vegetation Degradation	"The temporary or permanent reduction in the density, structure, species composition or productivity of vegetation cover (Conacher & Sala, 1998). Such as burn scars after natural brush fires or those ignited to clear the land for agriculture. These can also be new or old burn scars and seasonally dry or dead vegetation (CILSS, 2016).		1:38,000	✓	✓	✓	✓
6	Development	Settlements such as villages, towns, cities and local roads (CILSS, 2016).		1:38,000	✓	✓	✓	✓



Methodology

eCognition Visual Classification Key

	Classification	Description	Imagery Examples in Infrared	Scale	1988	2000	2010	2023
7	Roads	Major roads.		1:24,000	X	X	X	✓
8	Water Bodies	Areas with permanent or semi-permanent surface water such as the Gambia River and smaller waterways (CILSS, 2016).		1:38,000	✓	✓	✓	✓
9	Large-scale Mine	Open pit where gold is mined (CILSS, 2016).		1:38,000	X	X	X	✓

10	Mine Tailing Ponds	Structure or embankment that is built to retain gold mining waste or the byproduct of open pit mining such as fine-grained particles, waste water, arsenic and mercury (Adapted from Morrill, et al., 2020).		1:38,000	X	X	X	✓
11	Small-scale Artisanal Mine	Shallow-pocked mining locales usually near a water source where mineral extractions are mined from surface sand or gravel with little need for sophisticated tools (Allan, 2015).		1:15,000	X	✓	✓	✓
12	Intermediate-scale Artisanal Mine	The intermediate-scale gold mine that is not as expansive as a large-scale mine but also not as small as an ASGM mine either. The intermediate-scale mine has more infrastructure than an ASGM and is adjacent to a village or town		1:24,000	✓	X	✓	✓
13	Other	Unidentified class or pixels, shadows, smoke, clouds, reflections of clouds in the water.	To be decided as needed.		X	X	X	X
14	NoData	Frame border in eCognition.	To be decided as needed.		✓	✓	✓	✓
Class Count					8	9	10	13



Methodology

eCognition Workspace

MGIS_Cap_Workspace - Developer - [Dec_27_2023]

File View Image Objects Analysis Architect Classification Process Tools Window Help

20% main

View Settings

Object Levels

Image Layers

Image Layer	R	G	B
S2A_Blue			
S2A_Green			
S2A_Red			
S2A_NIR			
S2A_PCA_1			
S2A_PCA_2			
S2A_PCA_3			
S2A_PCA_4			
L8_Blue			
L8_Green			
L8_Red			
L8_NIR			
L8_SWIR_1			
L8_SWIR_2			
L8_PCA_1			
L8_PCA_2			
L8_PCA_3			
L8_PCA_4			
L8_PCA_5			
L8_PCA_6			
SN_Soils_Raster			
SN_StreamNetwork_Ra...			
DEM_SRTM			
NDVI_2023_TIF			
NDWI_2023_TIF			
SAVI_2023_TIF			

Object Level Settings

Object outline Outlines Draw Skeleton

Available Classes

Class	Color
unclassified	No Color
Closed-veget.	Green
Ecotone	Light Green
Open-vegetat.	Yellow-Green
Burn Scars	Brown
Development	Pink
Mine Tailing P.	Blue
Roads	Yellow
Water Bodies	Light Blue
Artisanal Sma.	Purple
NoData	White
Large-scale ...	Red

Process Tree

- Ruleset
 - PREPROCESSING
 - index layer NDVI 'NDVI_2023' (S2A_Red, S2A_NIR)
 - index layer NDWI 'NDWI_2023' (S2A_Green, S2A_NIR)
 - index layer SAVI 'SAVI_2023' (S2A_NIR, S2A_Red)
 - UNSUPERVISED CLASSIFICATION
 - unsupervised classification (ISODATA): [DEM_SRTM, S2A_B...
 - export image to to ExportImage
 - CREATE OR UPDATE GENERIC CLASSES
 - 15x: create/update class "<auto>"[-,superclass=-,group=-,r...
 - MULTI-THRESHOLD SEGMENTATION
 - 15x: multi-threshold: creating 'New Level': Class.104 <= 1 <
 - ASSIGN CLASSES
 - Class.106 at New Level: Closed-vegetation
 - Class.111 at New Level: Ecotone
 - Class.107 at New Level: Open-vegetation
 - Class.105 at New Level: NoData
 - Class.114 at New Level: Artisanal Small-scale Mine
 - Class.113 at New Level: Bare Soil
 - Class.110 at New Level: Water Bodies
 - Class.109 at New Level: Burn Scars
 - Class.112 at New Level: Development
 - Class.104 at New Level: Other
 - Class.108 at New Level: Roads
 - Class.115 at New Level: Large-scale Mine
 - Class.116 at New Level: Mine Tailing Ponds
 - Class.117, Class.118 at New Level: Intermediate-scale Artisa
 - REFINE
 - manual classification (brush: 5) Water Bodies -> Burn Scar
 - Artisanal Small-scale Mine, Bare Soil, Burn Scars, Closed-ve
 - Closed-vegetation with Area <= 10 Pxl at New Level: remo
 - Ecotone with Area <= 10 Pxl at New Level: remove objects
 - Open-vegetation with Area <= 10 Pxl at New Level: removi
 - Bare Soil with Area <= 10 Pxl at New Level: remove object
 - Burn Scars with Area <= 10 Pxl at New Level: remove objec
 - Development with Area <= 10 Pxl at New Level: remove ob
 - Roads with Area <= 10 Pxl at New Level: remove objects (r
 - Water Bodies with Area <= 10 Pxl at New Level: remove ob
 - Large-scale Mine with Area <= 10 Pxl at New Level: removi
 - Mine Tailing Ponds with Area <= 10 Pxl at New Level: remc
 - Artisanal Small-scale Mine with Area <= 10 Pxl at New Lev
 - Artisanal Small-scale Mine, Bare Soil, Burn Scars, Closed-ve
 - EXPORT
 - at New Level: export object shapes to SN_2023_Classificati

Process Properties

Auto name Class.106 at New Level

Setting

Setting	Value
Algorithm	assign class
Domain	
Scope	image object level
Level	New Level
Class filter	Class.106
Condition	---
Map	From Parent
Region	From Parent
Max. number of objects	all
Samples only	No
Algorithm parameters	
Use class	Closed-vegetation
Loops & cycles	
Loop while something change...	Yes
Number of cycles	1

Class Hierarchy

- classes
 - Artisanal Small-scale Mine
 - Bare Soil
 - Burn Scars
 - Class.104
 - Class.105
 - Class.106
 - Class.107
 - Class.108
 - Class.109
 - Class.110
 - Class.111
 - Class.112
 - Class.113
 - Class.114
 - Class.115
 - Class.116
 - Class.117
 - Class.118
 - Closed-vegetation
 - Development
 - Ecotone
 - Intermediate-scale Artisanal Mine
 - Large-scale Mine
 - Mine Tailing Ponds
 - NoData
 - Open-vegetation
 - Other
 - Roads
 - Water Bodies

Object Information Feature View

Object Levels

Image Layers

Point Clouds

Vector Layers

General Settings

Auto update

Apply to all views

Discard Apply

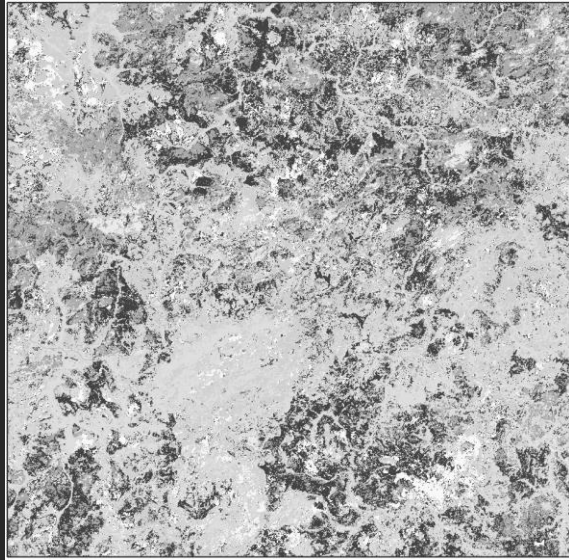
[130, 5845] = (756980.00, 1440265.00) Zoom:20%

RGB S2A_Blue linear (1.00%) 20% New Level/1 XY 167.977 Objects

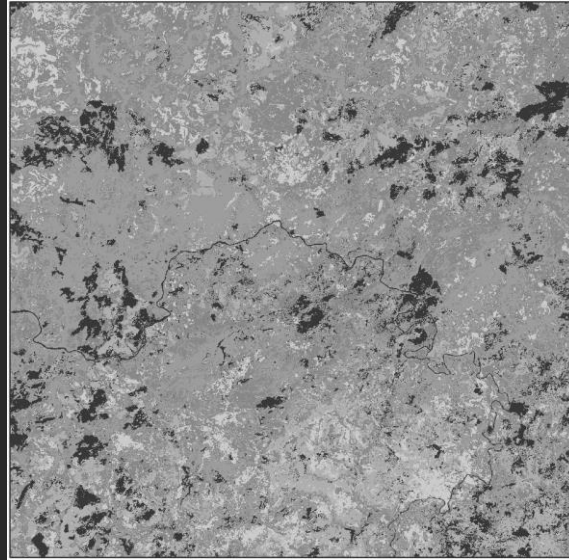


Methodology

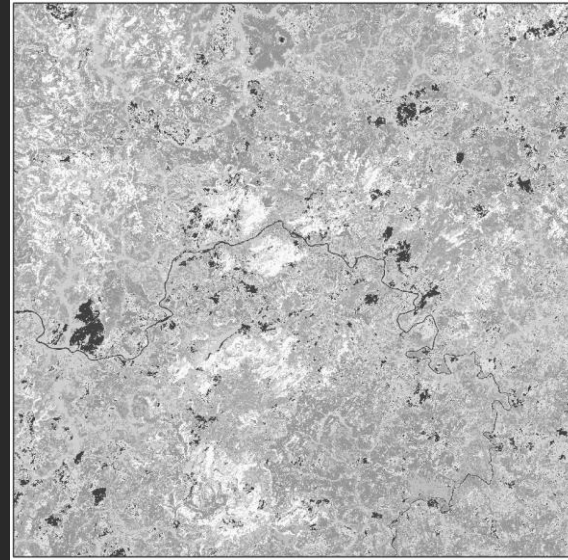
eCognition Unsupervised Classification
with Data Fusion for Each Year



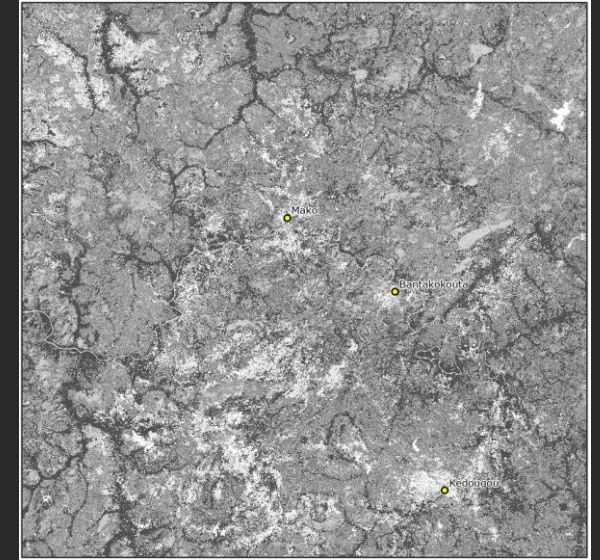
1988
30 meters



2000
15 meters



2010
30 meters

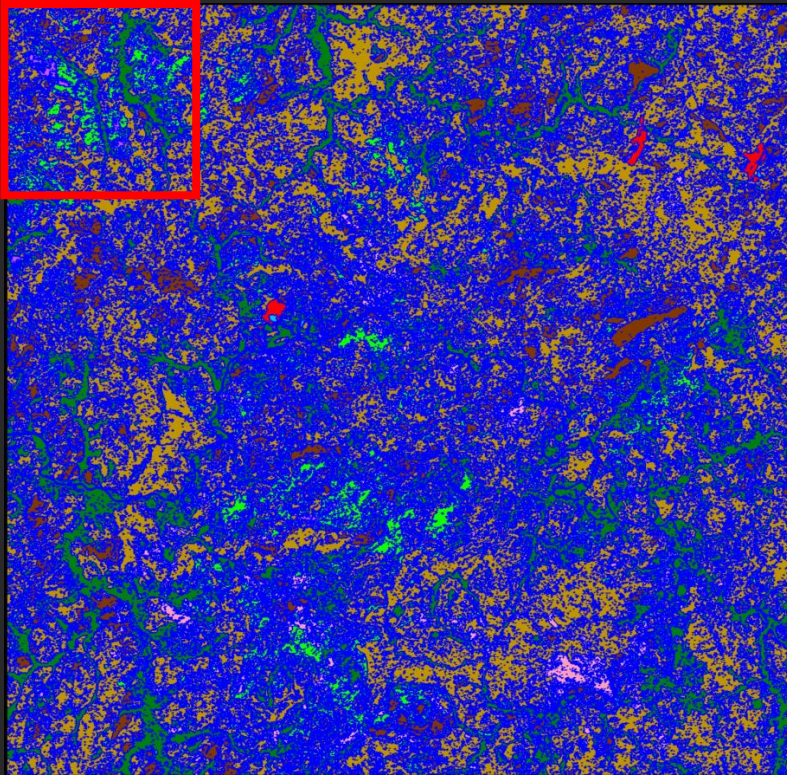


2023
10 meters

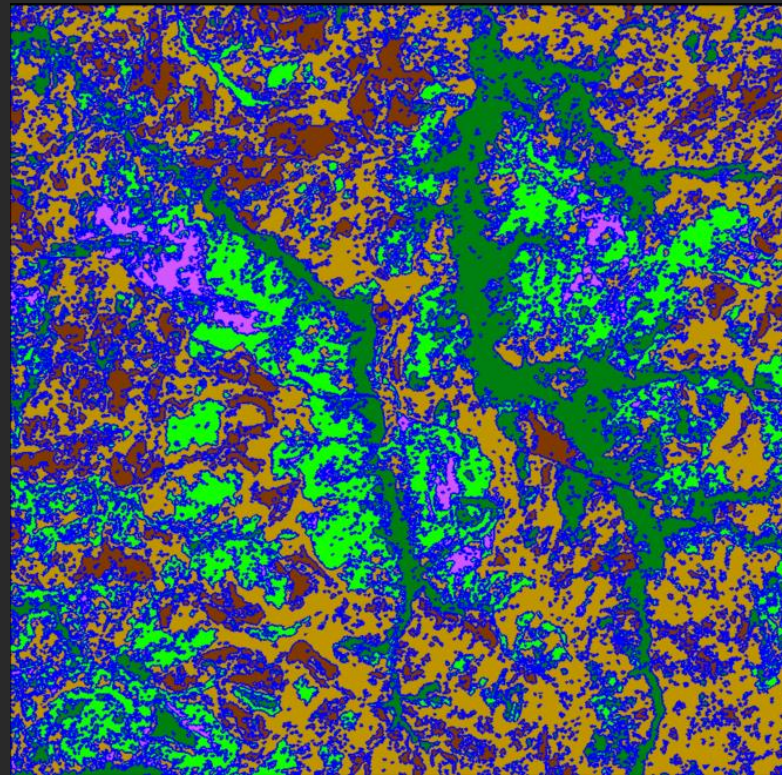


Methodology

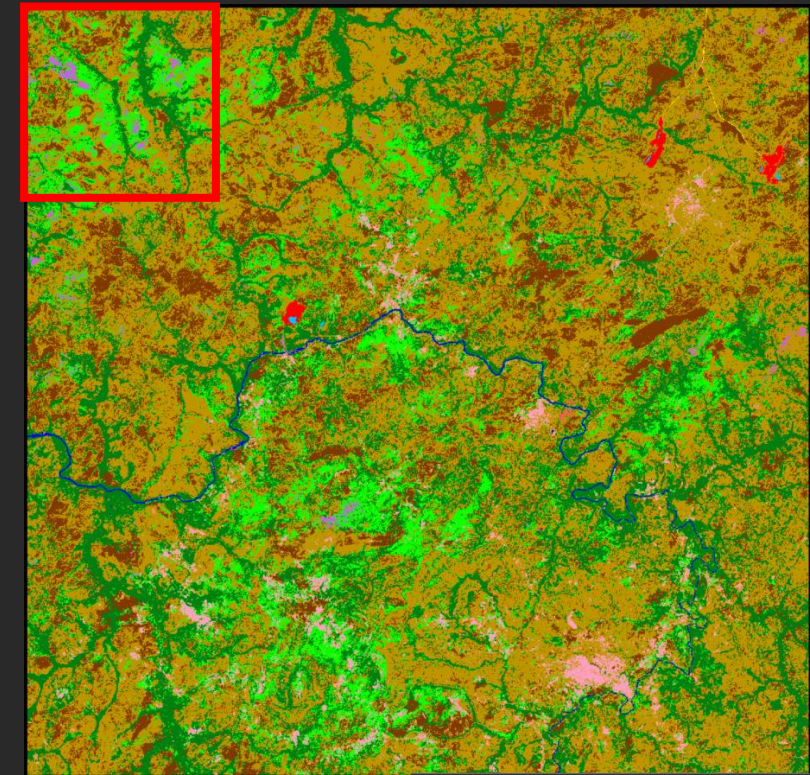
Example of the Results After Running the Multi-threshold Segmentation



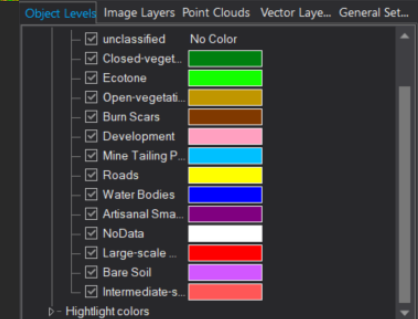
The vector results after running the multi-threshold segmentation algorithm in eCognition for 10-meter resolution (2023).



An up-close view of the segmentation results.



Results after classification and refinement steps.





Methodology

Post-processing in ArcGIS Pro

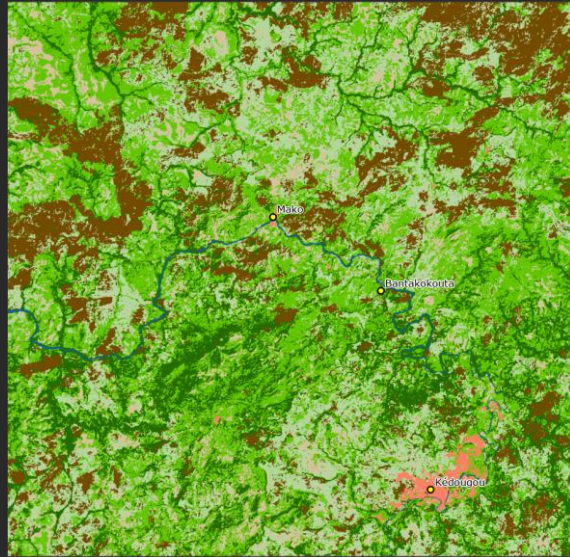
- Performed an **accuracy assessment** in ArcGIS Pro for the 2023 classification using stratified random sampling for 663 points.
- Performed a **change detection** analysis to track closed-vegetation (gallery forest) from 1988 to 2023.
- Last, the **presence-only prediction** tool in ArcGIS Pro was used to predict where the savanna chimpanzee habitat was suitable in 2023.





Results

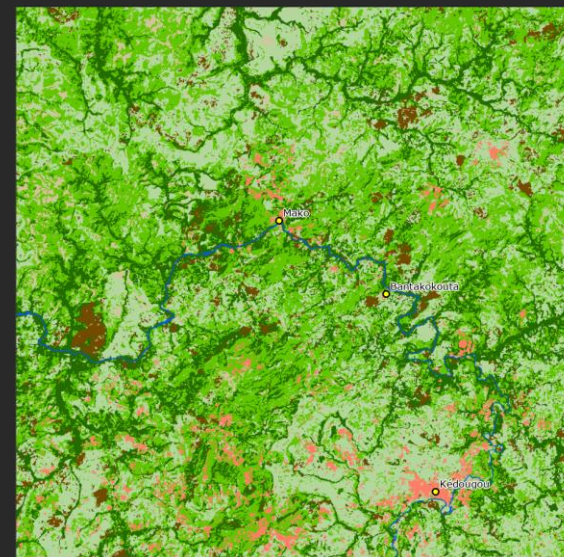
eCognition Classification for Area of Interest



1988



2000

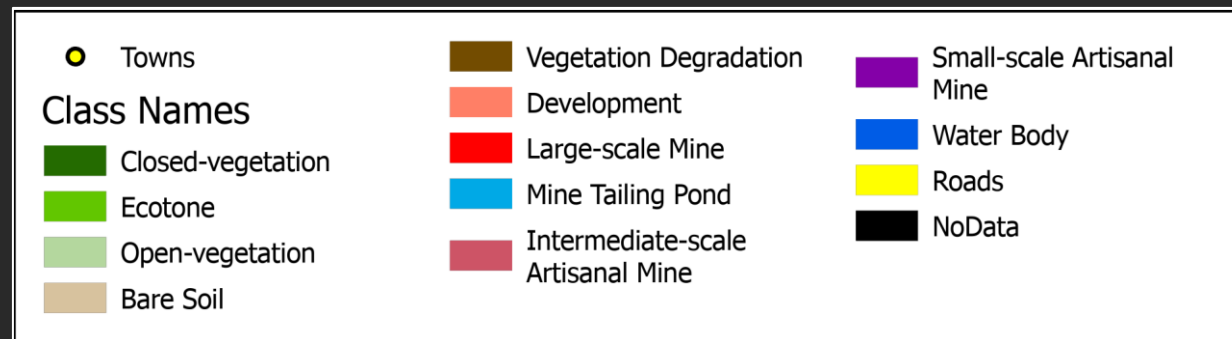


2010



2023

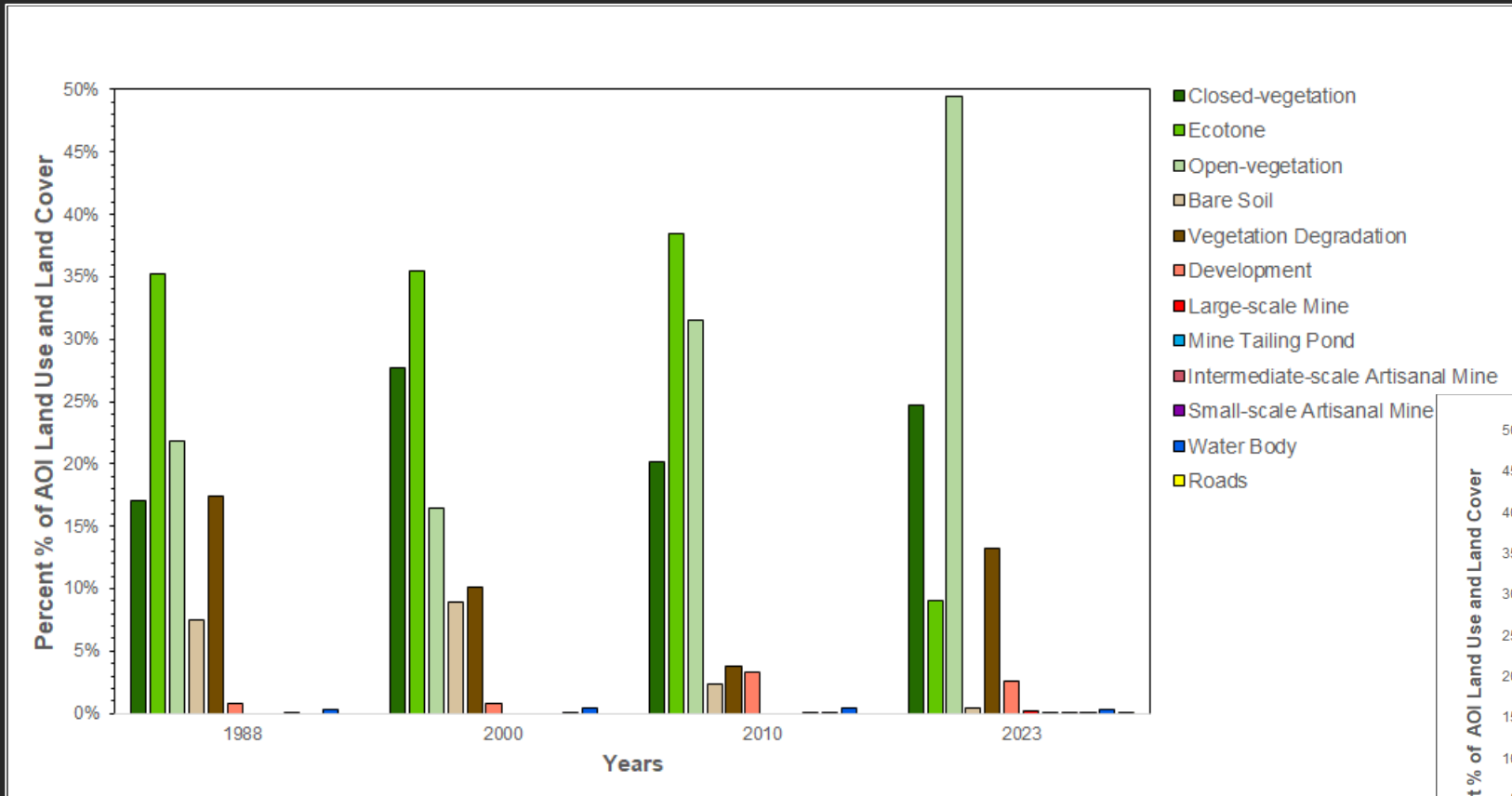
4409 km²



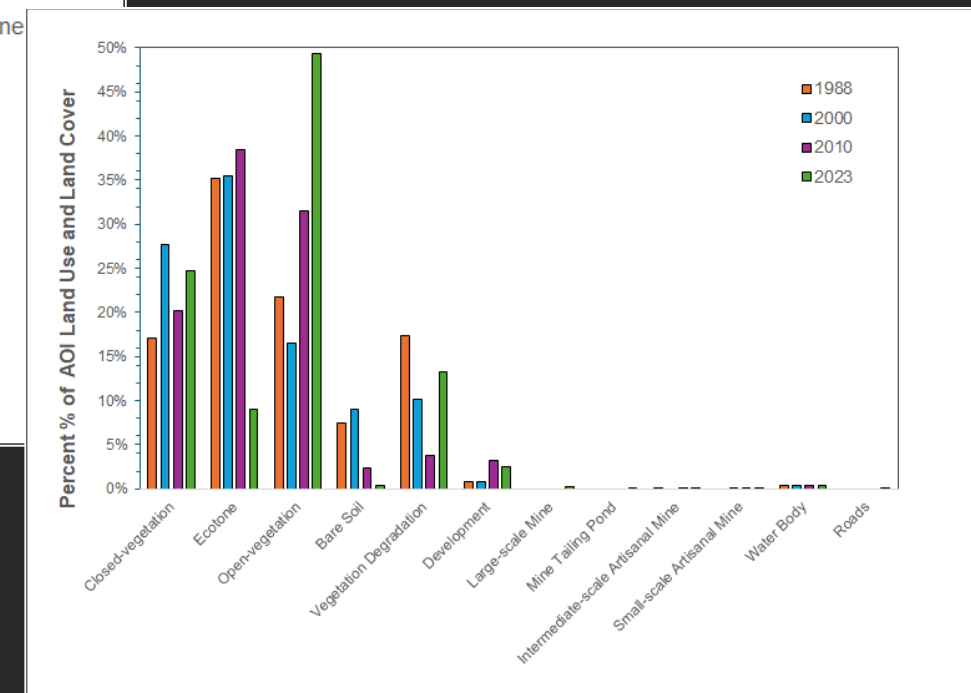


Results

eCognition Classification for Area of Interest



- From 1988 to 2023, the LULC has shifted to predominately open-vegetation. Ecotone increased from 1988 to 2010; however, in only 13 years, from 2010 to 2023, the ecotone has dramatically declined, and closed-vegetation (Gallery Forests) has increased.

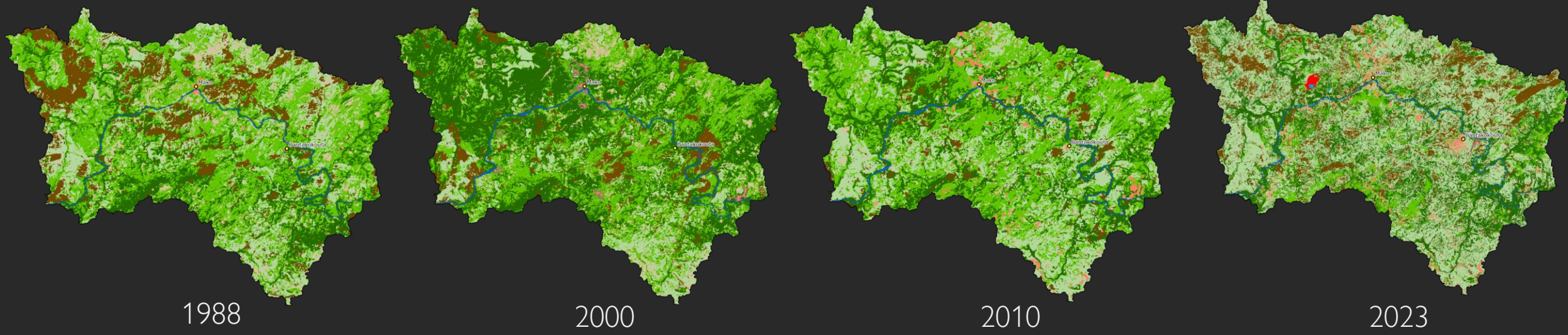


- This time interval from 2010 to 2023 corresponds with the gold mining boom in the study area.

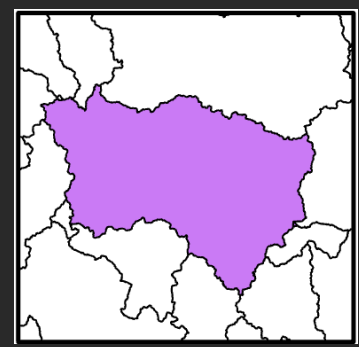
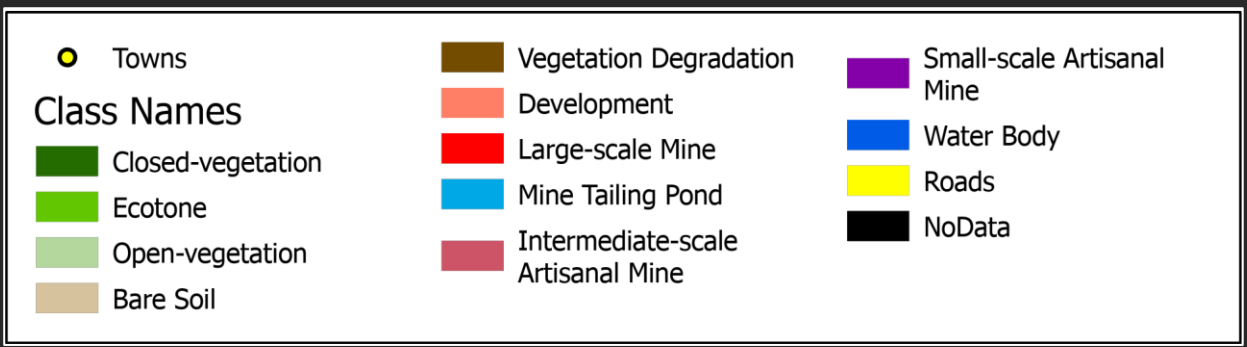


Results

eCognition Classification for Subregional Watershed



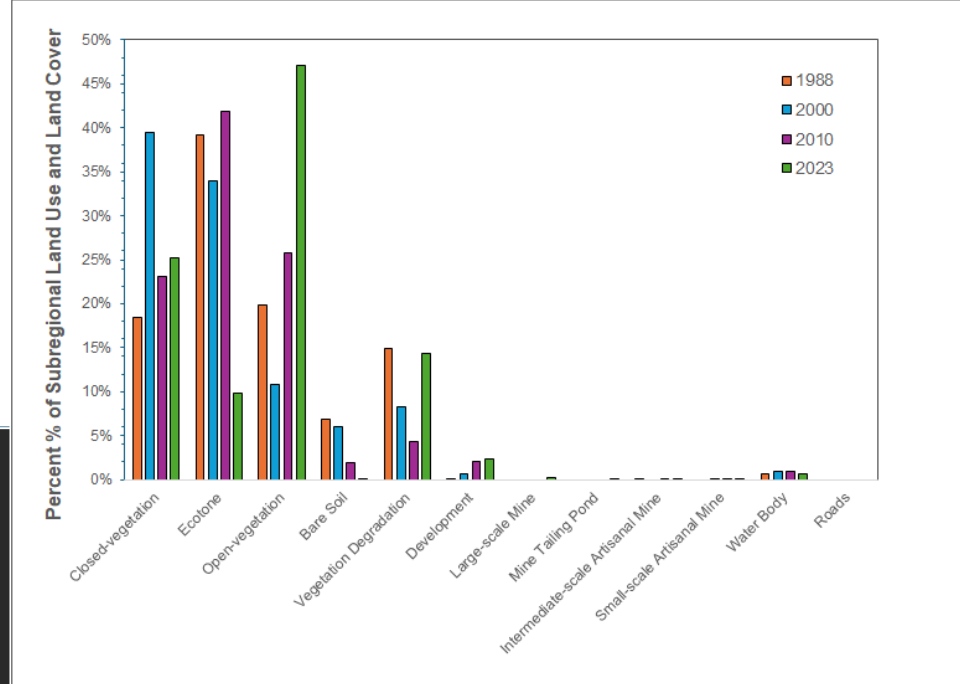
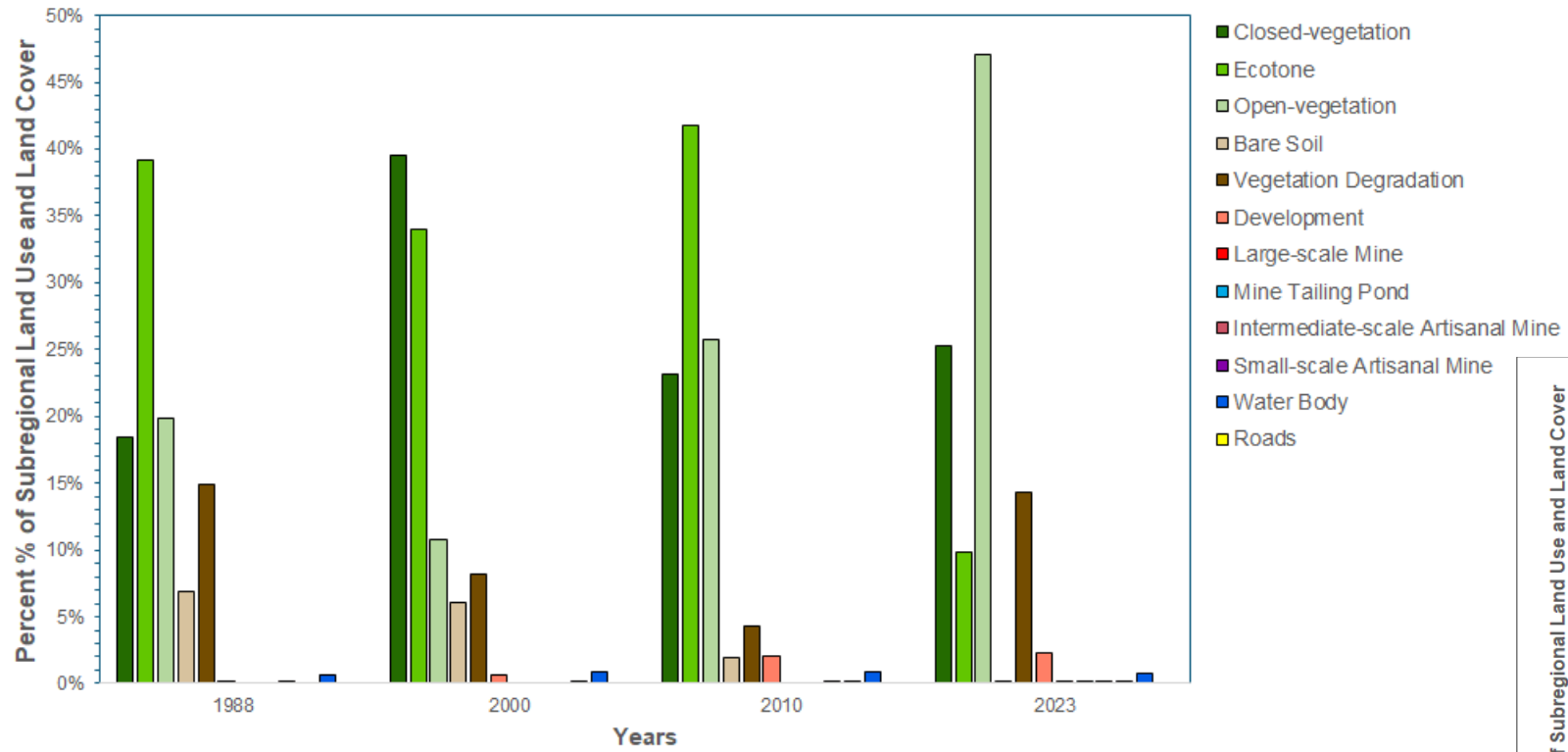
1291 km²





Results

eCognition Classification for Subregional Watershed



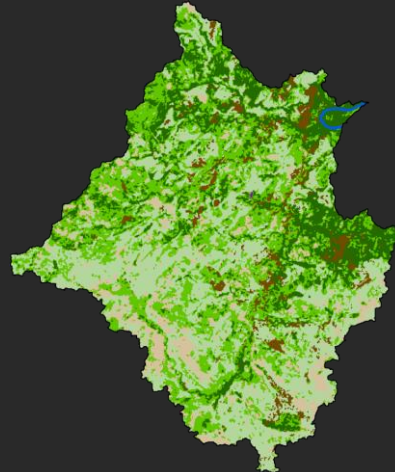


Results

eCognition Classification for Local Watershed (this is Fongoli)



1988



2000

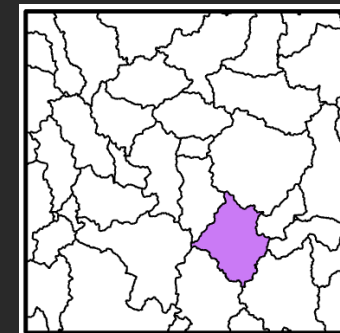
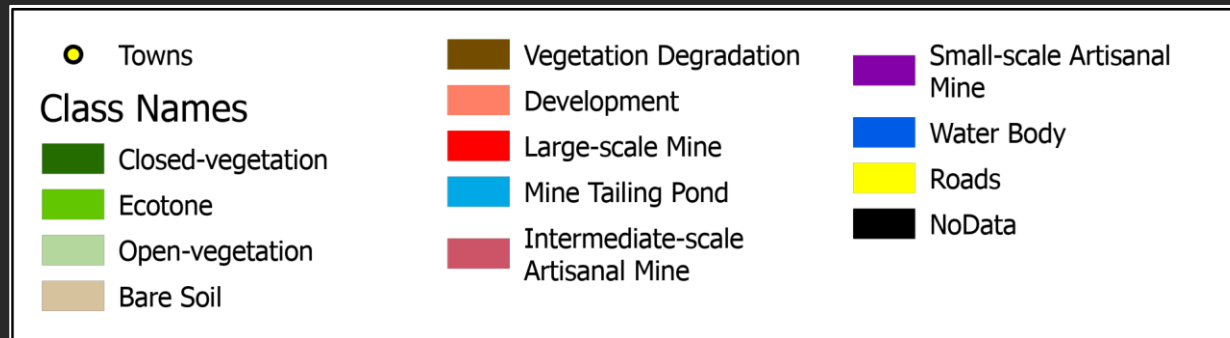


2010



2023

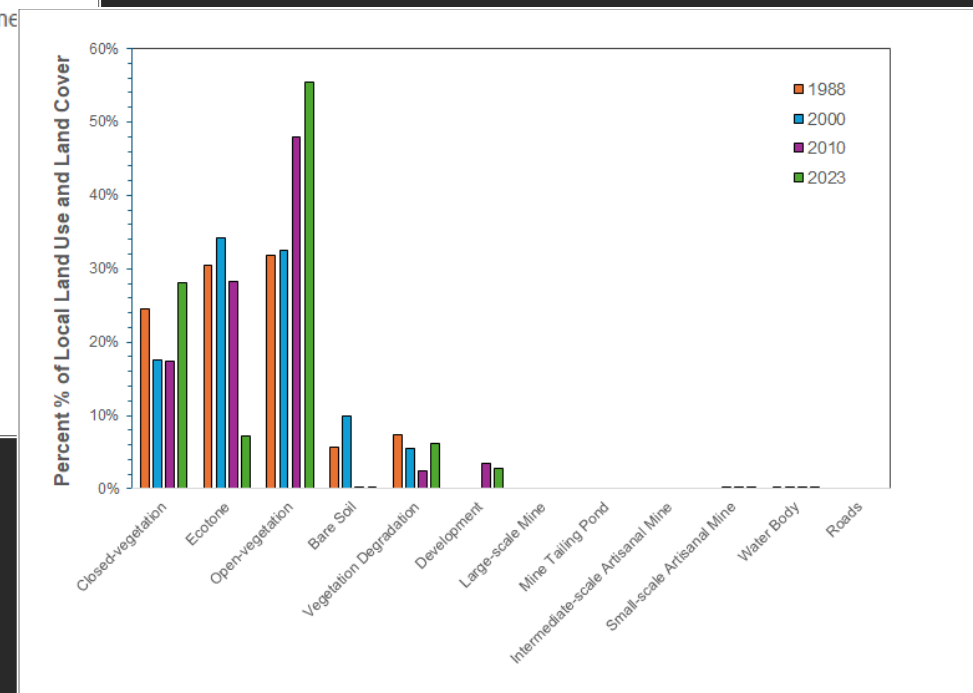
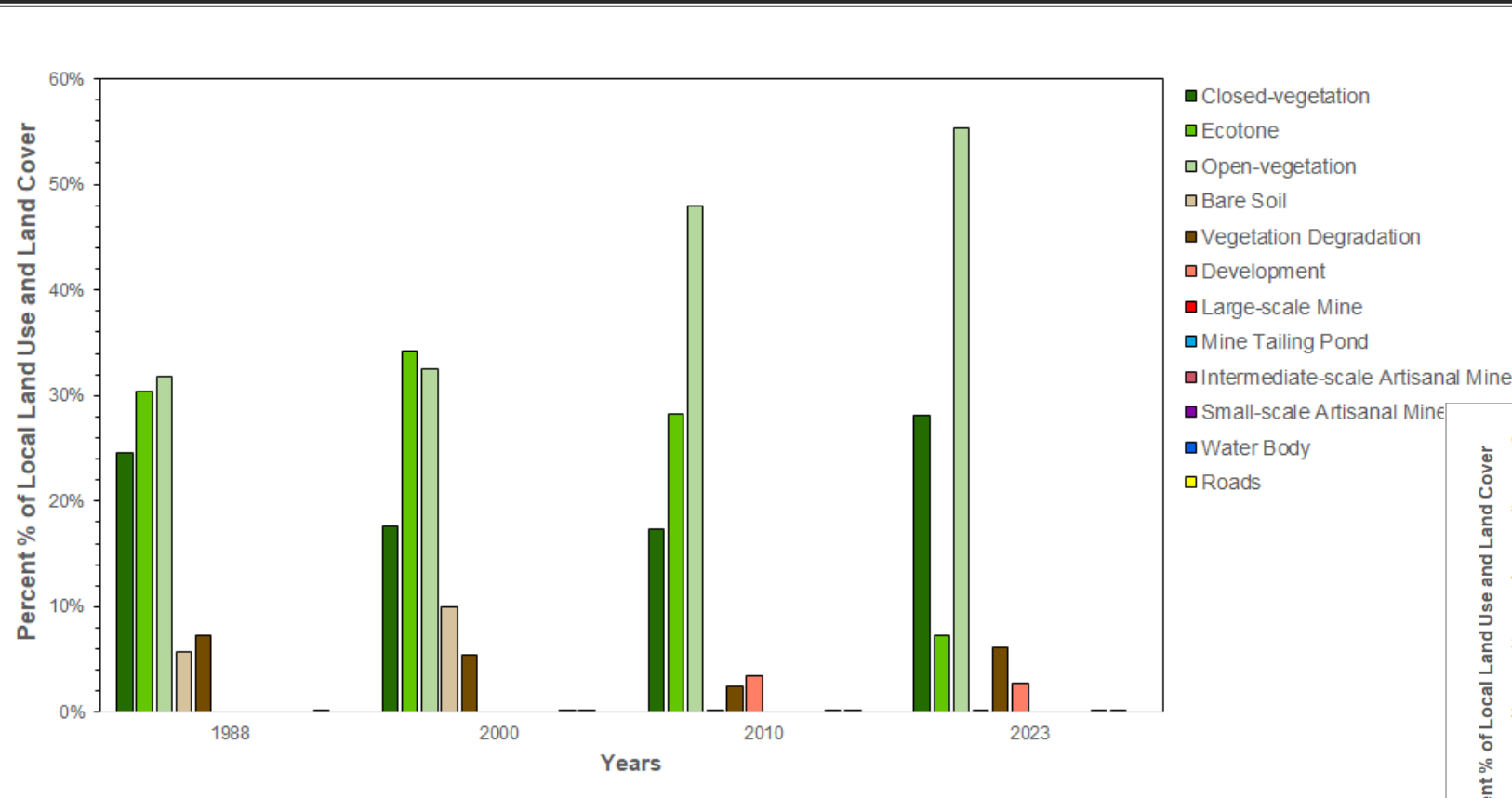
157 km²





Results

eCognition Classification for Local Watershed





Results

ArcGIS Pro Accuracy Assessment for 2023

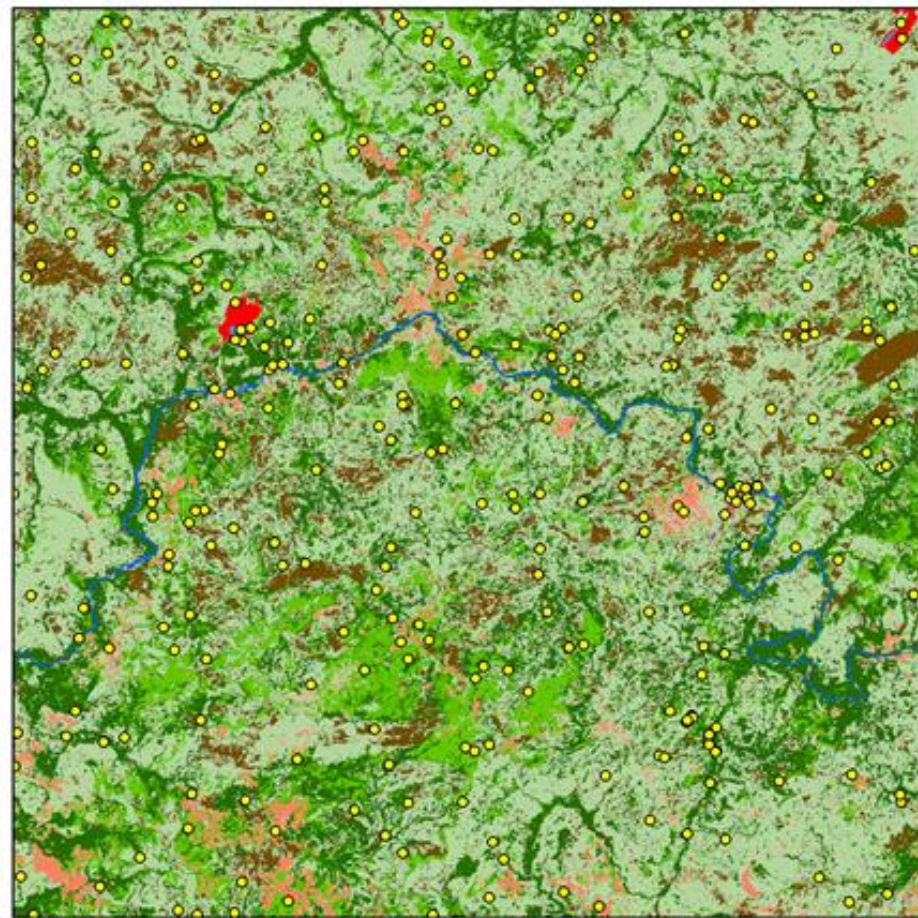
663 points using stratified random sampling and PlanetScope (4.77 meters) as the reference data for the same day and year.

		Reference Data (User's Data)															
Map Classified Data (Producer's Data)	Class For 2023	Small-scale Artisanal Mine	Bare Soil	Vegetation Degradation	Closed-vegetation	Development	Ecotone	Intermediate-scale Artisanal Mine	Large-scale Mine	Mine Tailing Pond	Open-vegetation	Roads	Water Body	Total	User's Accuracy		
	Small-scale Artisanal Mine	10	0	0	0	0	0	0	0	0	0	0	0	0	10	100%	
	Bare Soil	0	10	0	0	0	0	0	0	0	0	0	0	0	10	100%	
	Vegetation Degradation	0	1	74	1	0	1	0	0	0	3	0	0	80	93%		
	Closed-vegetation	0	1	1	93	3	44	0	0	0	6	0	0	148	63%		
	Development	0	3	0	1	11	0	0	0	0	0	0	0	15	73%		
	Ecotone	0	4	0	8	2	38	0	0	0	2	0	0	54	70%		
	Intermediate-scale Artisanal Mine	0	0	0	0	0	0	10	0	0	0	0	0	10	100%		
	Large-scale Mine	0	0	0	0	0	0	0	10	0	0	0	0	10	100%		
	Mine Tailing Pond	0	0	0	0	1	0	0	0	9	0	0	0	10	90%		
	Open-vegetation	0	3	12	8	3	14	1	0	0	254	1	0	296	86%		
	Roads	0	0	0	0	0	2	0	0	0	2	6	0	10	60%		
	Water Body	0	0	1	1	0	0	0	0	0	0	0	8	10	80%		
	Total	10	22	88	112	20	99	11	10	9	267	7	8	663			
Producer's Accuracy	100%	45%	84%	83%	55%	38%	91%	100%	100%	95%	86%	100%		Overall Accuracy	80%	Kappa	
															80%	74%	



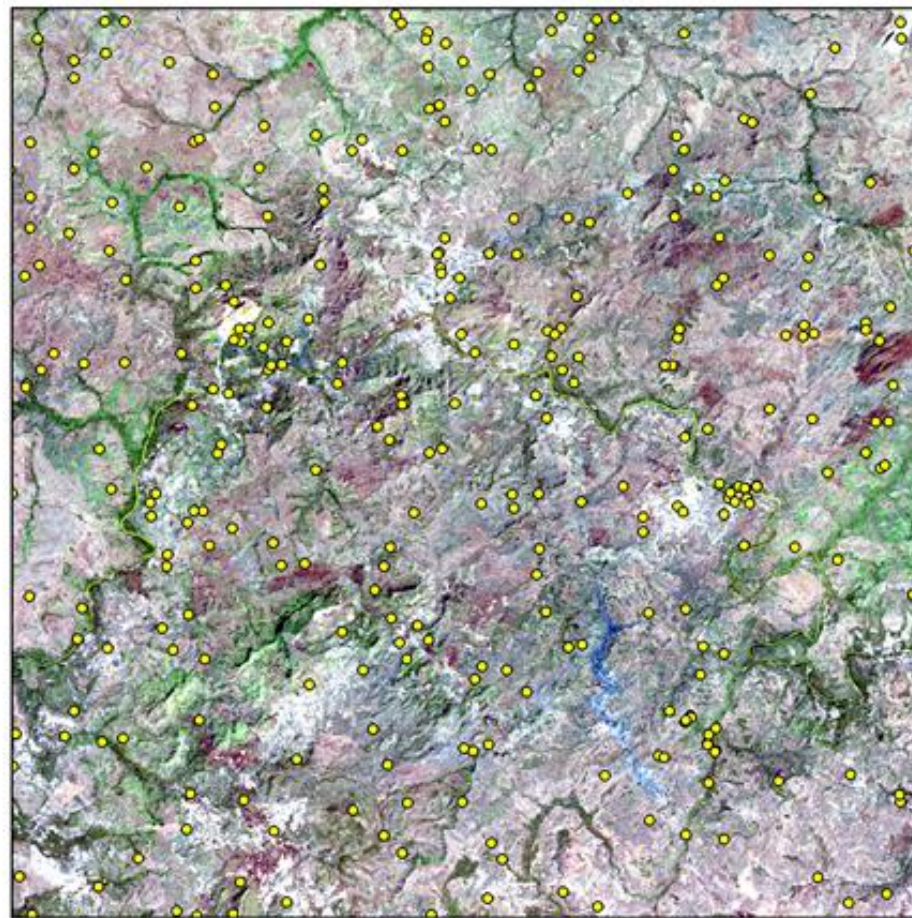
Results

ArcGIS Pro Accuracy Assessment for 2023



0 5 10 20 Kilometers

Producer's Map



User's Map



The distribution of 663 random points on the producer's map (left) and the user's reference data (right).

The producer's map is 10 meter resolution (Sentinel 2A).

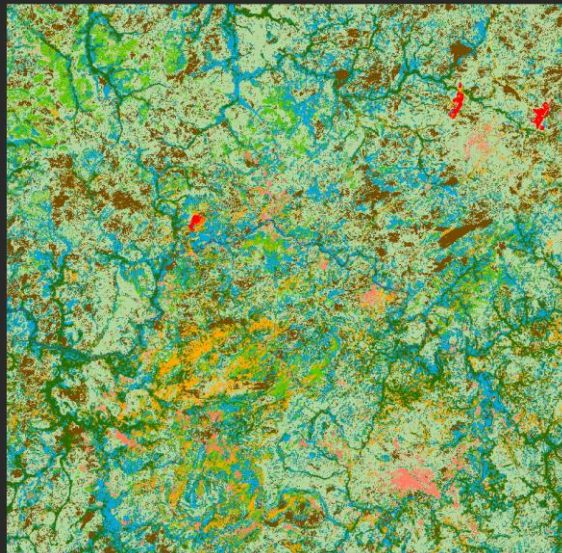
The user's map is PlanetScope (NICFI) 4.77 meter resolution.

Both are on the same day, December 27, 2023.

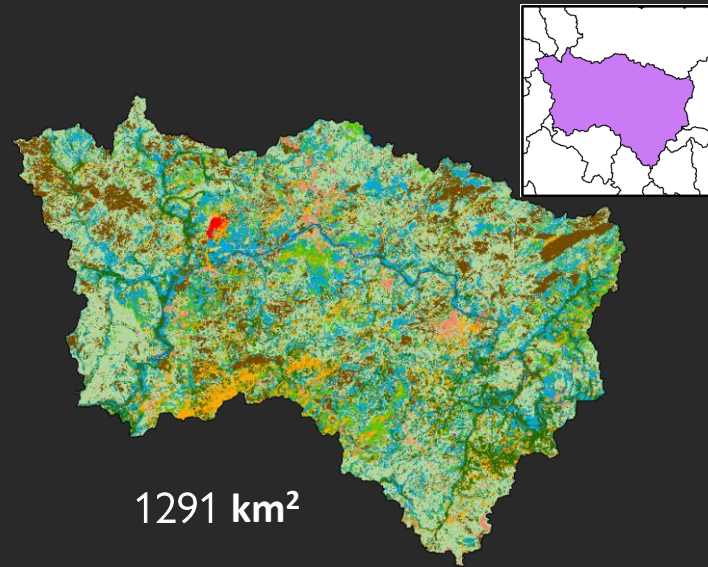


Results

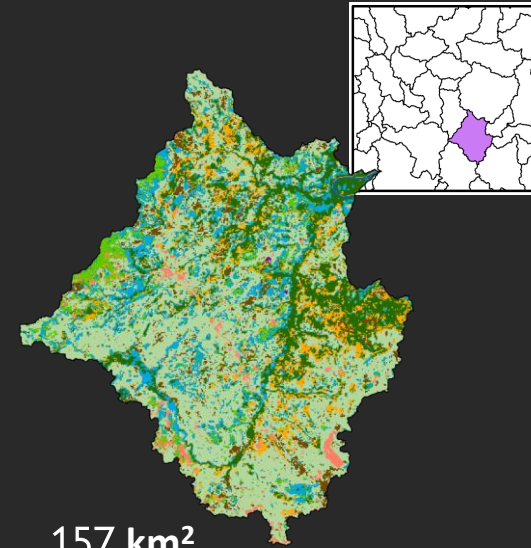
ArcGIS Pro Change Detection for 1988 to 2023



4409 km²



1291 km²



157 km²

Change Designation

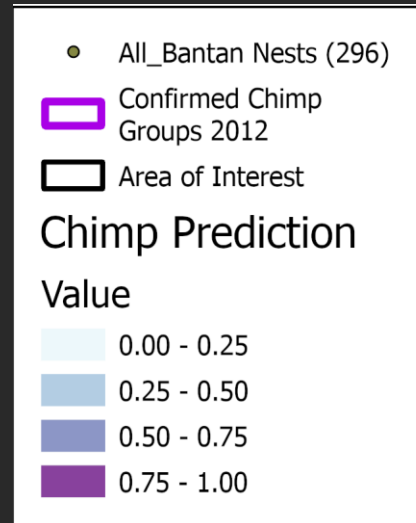
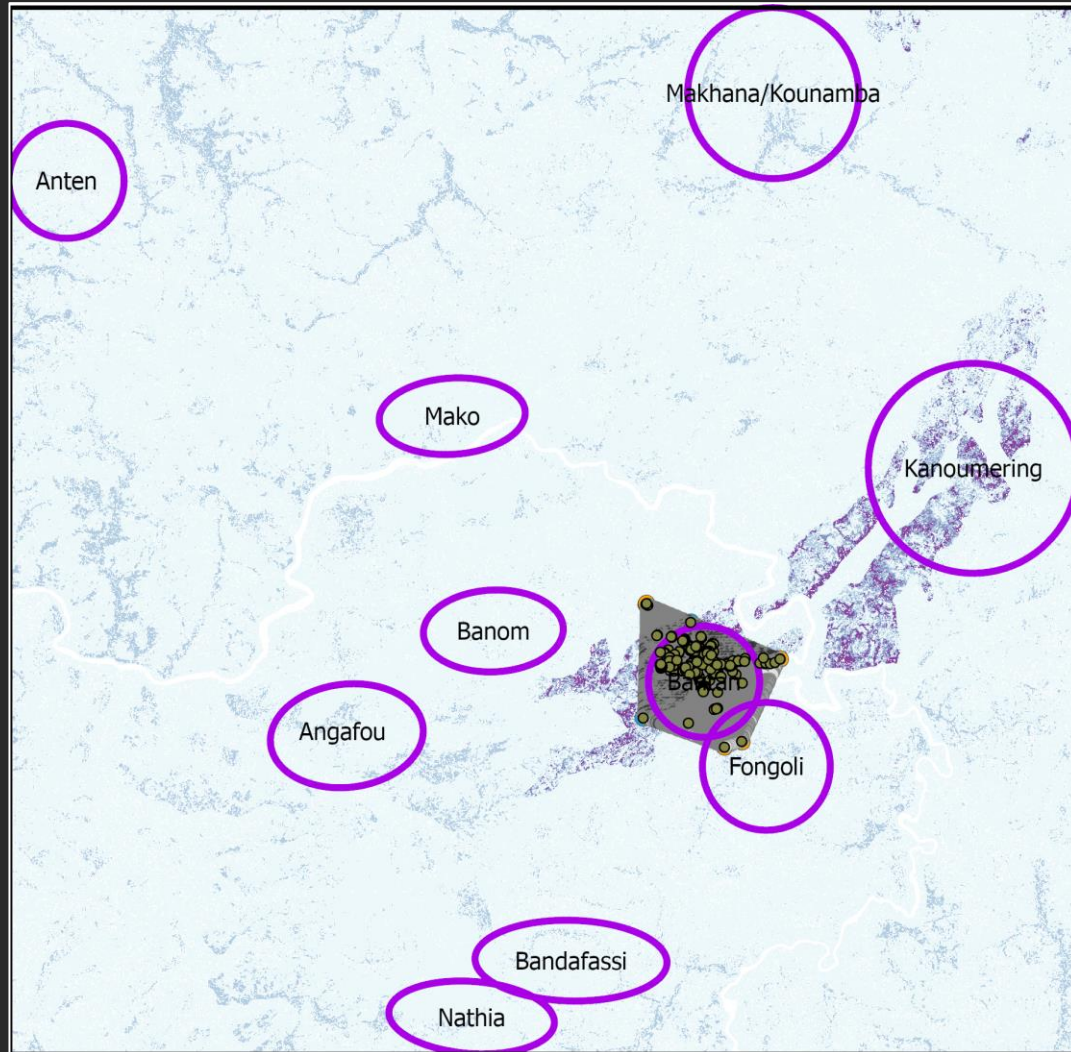
- Bare Soil->Closed-vegetation
- Closed-vegetation->Bare Soil
- Closed-vegetation->Closed-vegetation
- Closed-vegetation->Development
- Closed-vegetation->Ecotone
- Closed-vegetation->Intermediate-scale Artisanal Mine
- Closed-vegetation->Large-scale Mine
- Closed-vegetation->Mine Tailing Ponds
- Closed-vegetation->Open-vegetation
- Closed-vegetation->Roads
- Closed-vegetation->Small-scale Artisanal Mine
- Closed-vegetation->Water Body
- Ecotone->Closed-vegetation
- Intermediate-Scale Artisanal Mine->Closed-vegetation
- Open-vegetation->Closed-vegetation
- Vegetation Degradation->Closed-vegetation
- Water Body->Closed-vegetation

Closed-vegetation Survey Scale	A No Change from 1988 to 2023 Color Code: Forest Green	B LULC from 1988 to Closed-vegetation in 2023 (Added) Color Code: Morea Blue	C Closed-vegetation from 1988 to LULC in 2023 (Subtracted) Color Code: Electron Gold	A + B = 2023 Total km ²	A + C = 1988 Total km ²	Closed-vegetation Gain/Loss from 1988 to 2023
AOI Change from 1988 to 2023	443	642	267	1,086	710	376
Subregional Change from 1988 to 2023	121	205	102	327	223	103
Local Change from 1988 to 2023	24	21	14	44	37	7



Results

ArcGIS Pro Presence-only Prediction for 2023

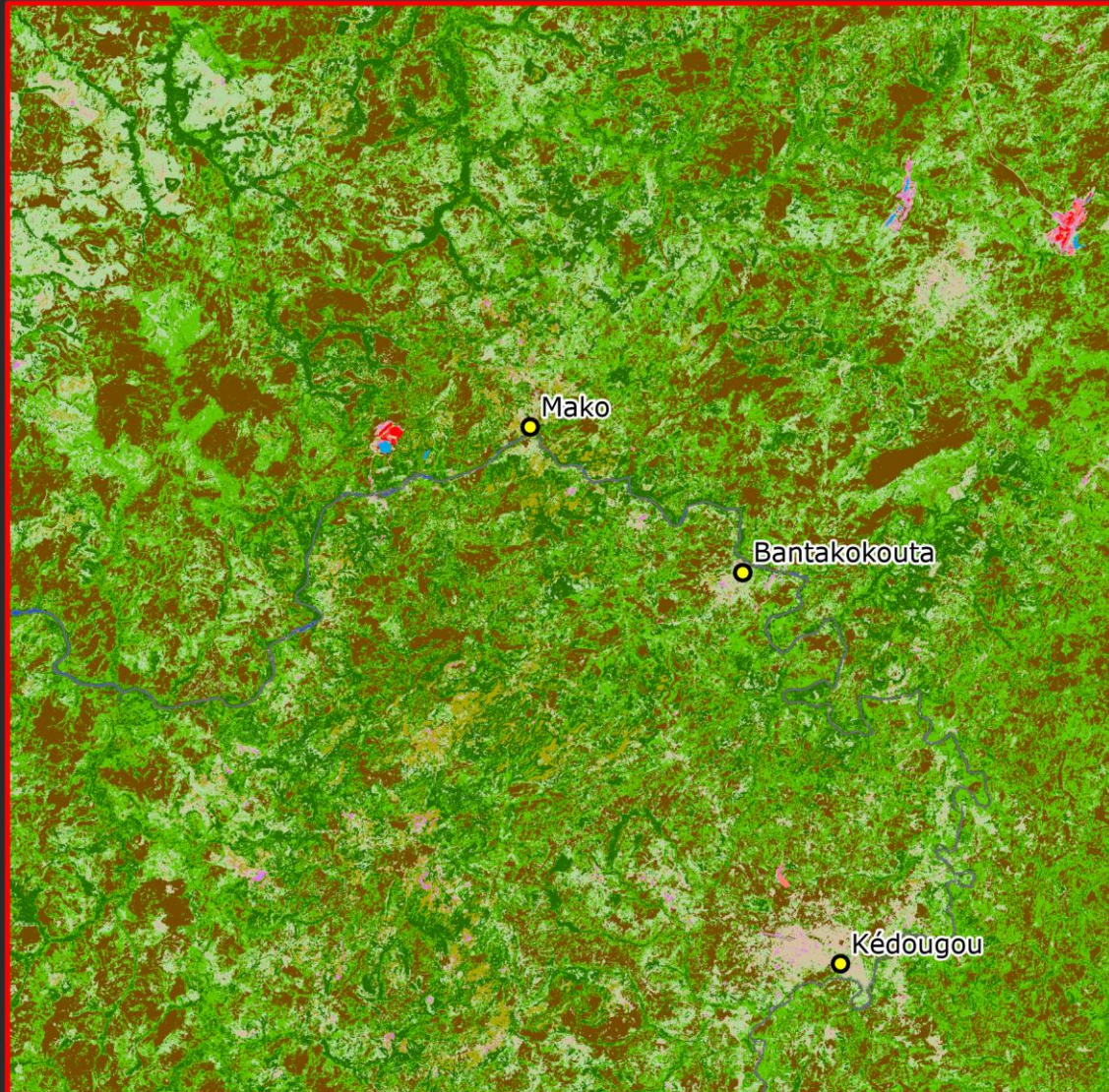


- AUC = 0.8242, Omission Rate = 0.2748
- Applied 296 opportunistic chimpanzee nesting points, and the model used 262, and 190 classified as presence.
- Used the 2023 classified imagery, along with the 3 different soils tables, elevation, aspect, slope, the stream network, and roads from OpenStreetMap.
- It is interesting that it started to include the gallery forests surrounding the nest point cluster. If the points were distributed throughout the area, it may have shown more habitat.

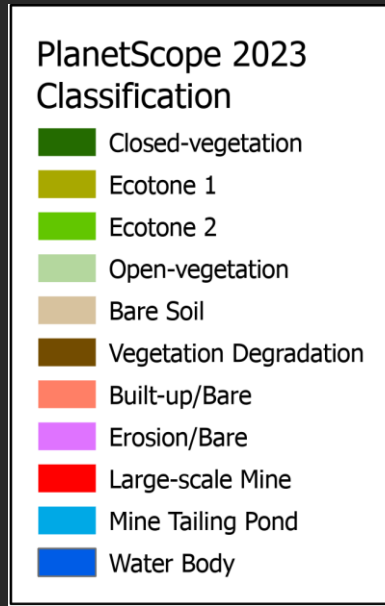


Results

Extra: PlanetScope Classification



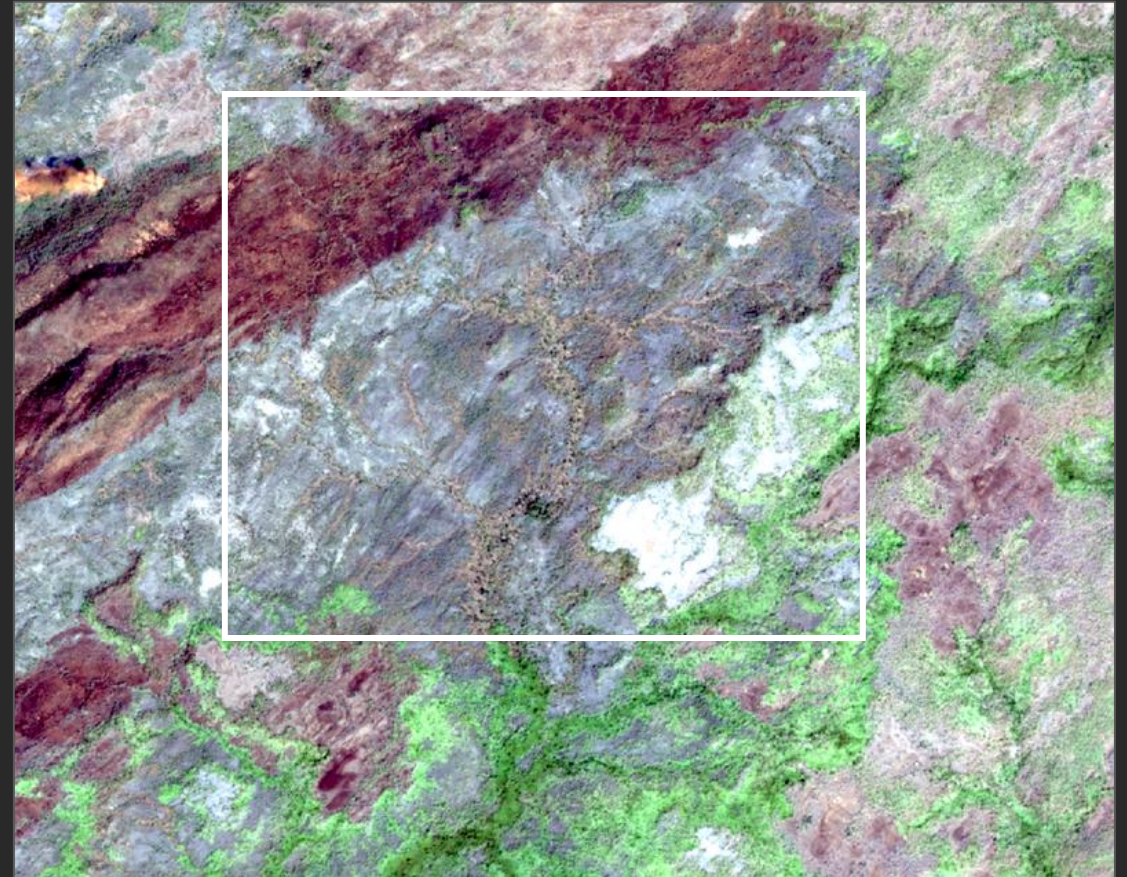
- The resolution for PlanetScope is 4.77 meters, and it took about 4 hours to run the unsupervised classification.
- But there is a lot more detail in the classification.



Senegal Land Cover Classes
Agriculture
Agriculture in shallows and recession
Bare soil
Bowe
Forest
Gallery forest and riparian forest
Irrigated agriculture
Mangrove
Open Mine
Plantation
Rocky land
Sahelian short grass savanna
Sandy Area
Savanna
Settlements
Steppe
Swamp Forest
Thicket
Water Bodies
Wetland - floodplain
Woodland

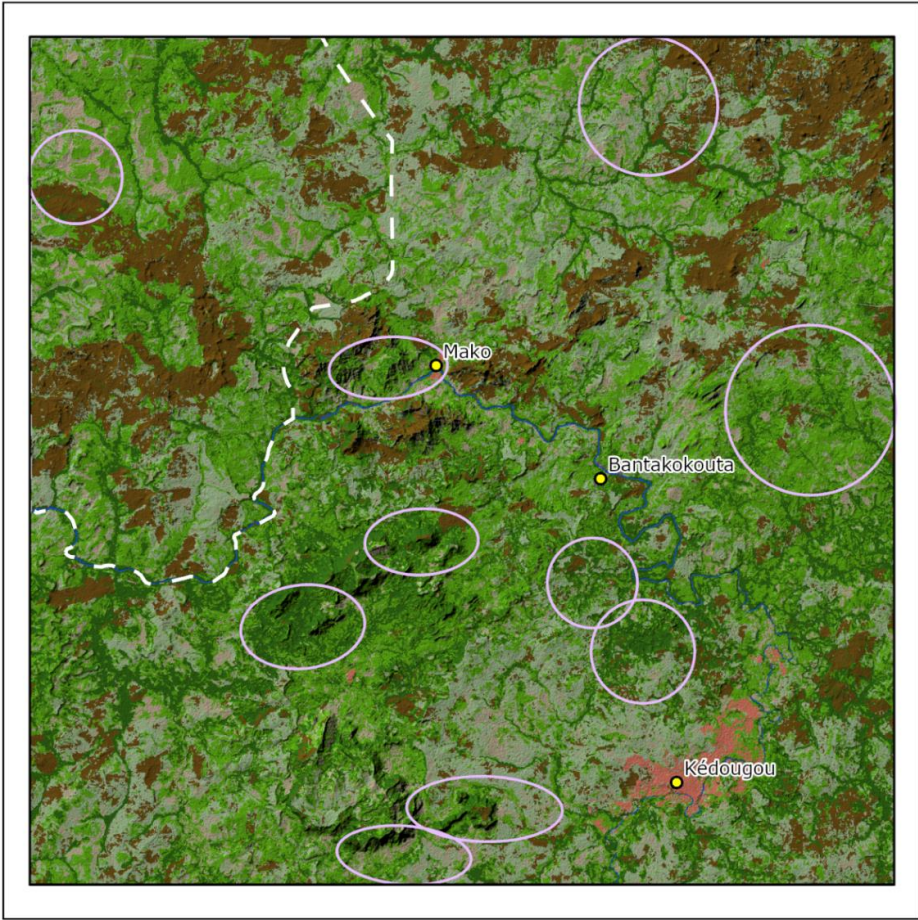


The new extension of the Sabodala mine north of the area of interest in this project. The picture depicts the mining infrastructure cutting through a gallery forest in two locations and the position of the mine tailing ponds next to the gallery forest watercourse (Photo credit to Google Earth).



PlanetScope (4.77 meters), December 27, 2023, natural bands (RGB 321). Showing an unhealthy, likely burned, or degrading gallery forest branch.

2023 Classification	
Class Names	
■ Closed-vegetation	■ Intermediate-scale Artisanal Mine
■ Ecotone	■ Small-scale Artisanal Mine
■ Open-vegetation	■ Water Body
■ Bare Soil	■ Roads
■ Vegetation Degradation	● Towns
■ Development	□ Area of Interest
■ Large-scale Mine	□ Niokolo-Koba National Park
■ Mine Tailing Pond	□ 2012 Confirmed Chimpanzee Groups



0 5 10 20 Kilometers

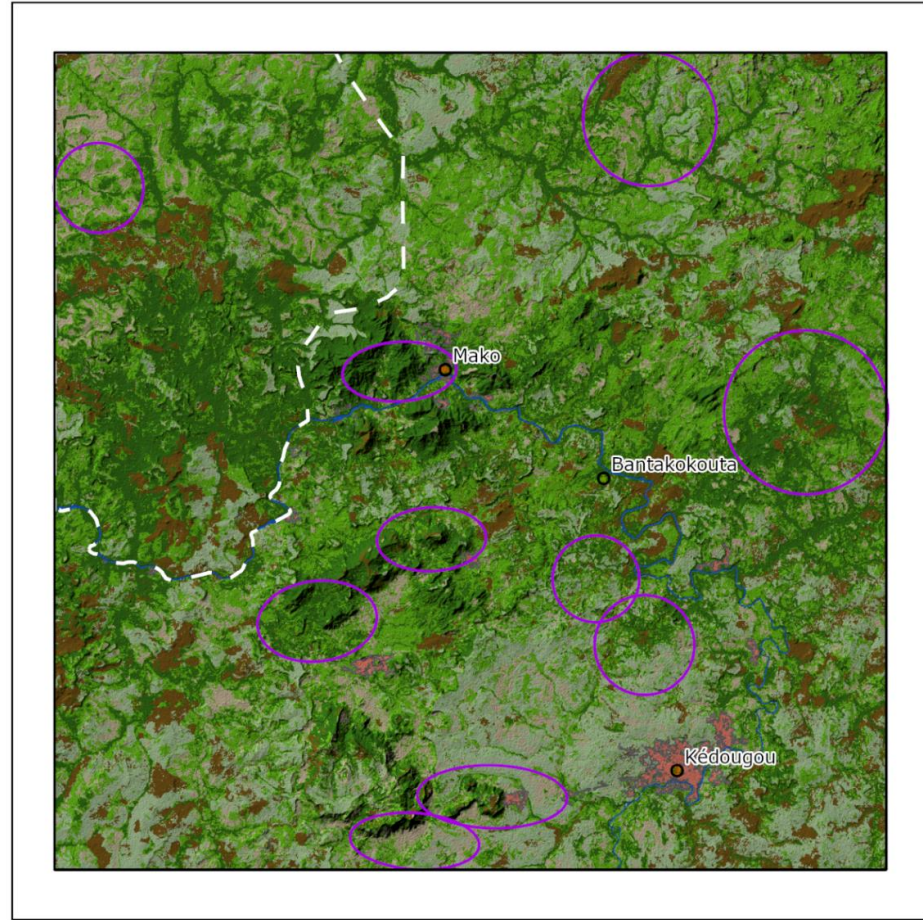


1988 Classification

Class Names

- Closed-vegetation
- Ecotone
- Open-vegetation
- Bare Soil
- Vegetation Degradation
- Development
- Intermediate-scale Mine
- Water Body
- Towns
- Area of Interest
- Niokolo-Koba National Park
- 2012 Confirmed Chimpanzee Groups

Constructed by
 Claudette Sandoval-Green
 28 April 2024
 30 meters, Landsat 5 TM, C2L2
 Unsupervised Classification
 Trimble eCognition
 Kedougou, Senegal
 WGS 1984 UTM Zone 28 N



0 5 10 20 Kilometers



2000 Classification

Class Names

- Closed-vegetation
- Ecotone
- Open-vegetation
- Bare Soil
- Vegetation Degradation
- Development
- Small-scale Artisanal Mine
- Water Body
- Towns
- Area of Interest
- Niokolo-Koba National Park
- 2012 Confirmed Chimpanzee Groups

Constructed by
 Claudette Sandoval-Green
 28 April 2024
 15 meters, Landsat 7 ETM, C2L2
 Unsupervised Classification
 Trimble eCognition
 Kedougou, Senegal
 WGS 1984 UTM Zone 28 N

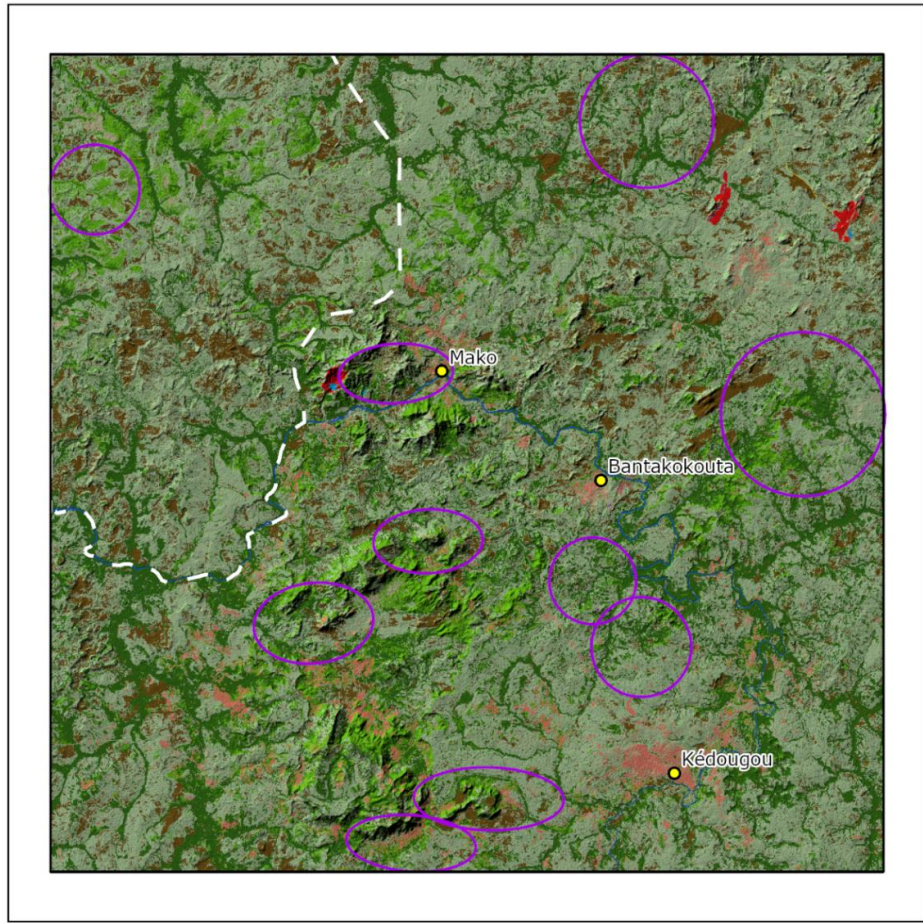


0 5 10 20 Kilometers



2010 Classification	
Class Names	
	Closed-vegetation
	Ecotone
	Open-vegetation
	Bare Soil
	Vegetation Degradation
	Development
	Intermediate-scale Artisanal Mine
	Small-scale Artisanal Mine
	Water Bodies
	Towns
	Area of Interest
	Niokolo-Koba National Park
	2012 Confirmed Chimpanzee Groups

Constructed by
 Claudette Sandoval-Green
 28 April 2024
 30 meters, Landsat 5 TM, C2L2
 Unsupervised Classification
 Trimble eCognition
 Kédougou, Senegal
 WGS 1984 UTM Zone 28 N



0 5 10 20 Kilometers



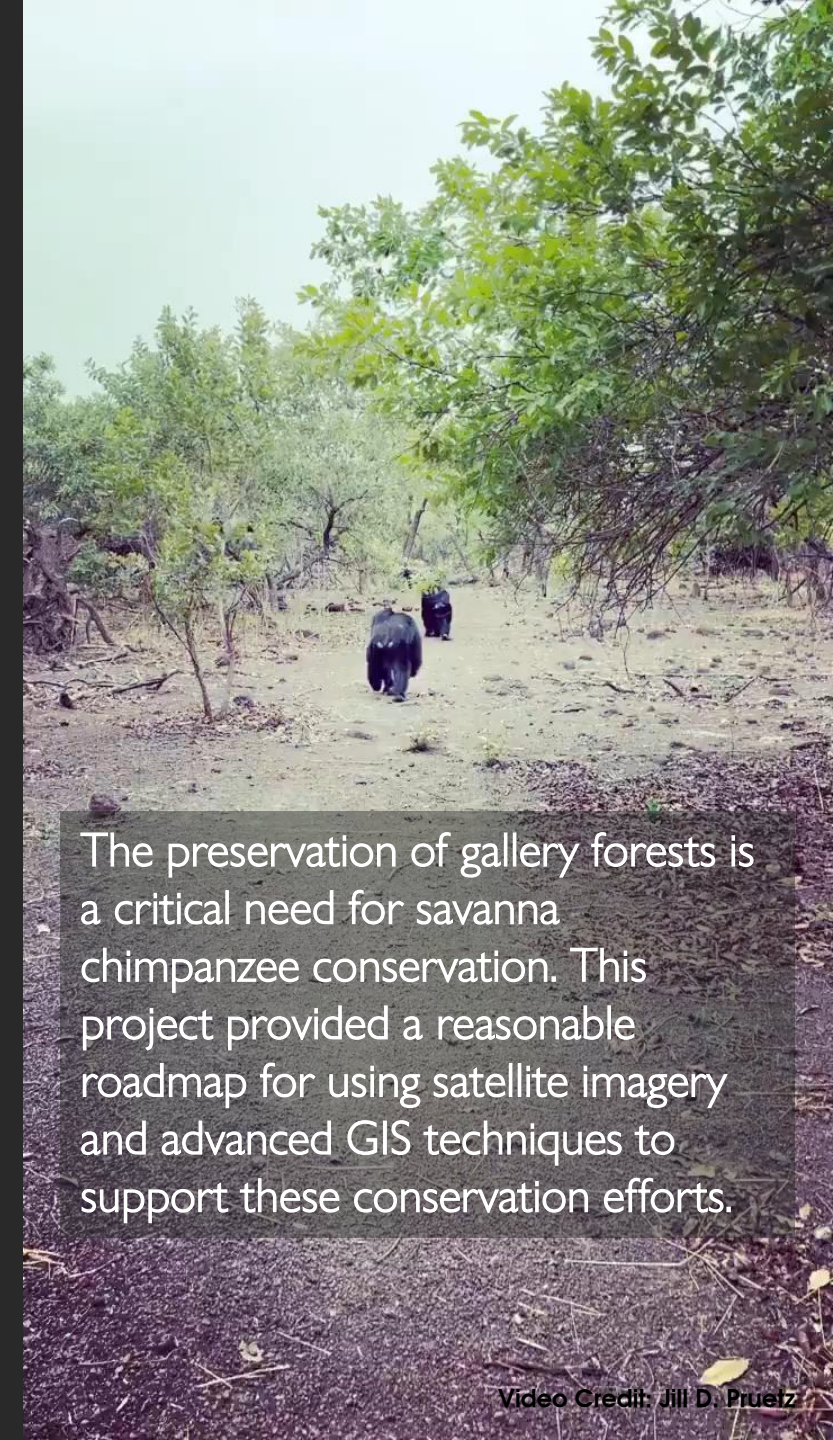
2023 Classification	
Class Names	
	Closed-vegetation
	Ecotone
	Open-vegetation
	Bare Soil
	Vegetation Degradation
	Development
	Large-scale Mine
	Mine Tailing Pond
	Intermediate-scale Artisanal Mine
	Small-scale Artisanal Mine
	Water Body
	Roads
	Towns
	Area of Interest
	Niokolo-Koba National Park
	2012 Confirmed Chimpanzee Groups

Constructed by
 Claudette Sandoval-Green
 28 April 2024
 10 meters, Sentinel 2A
 Unsupervised Classification
 Trimble eCognition
 Kédougou, Senegal
 WGS 1984 UTM Zone 28 N



Lessons Learned

- It's important to remember that all chimpanzees, including the savanna chimpanzee, are dependent on trees for their habitat. This fundamental fact should guide research and conservation efforts.
- The unsupervised classification results were very encouraging. The gallery forests were always the most accessible and most robust cluster to classify. Ecotone is harder.
- Classification should be a two-person job, at the very least, that includes the remote sensing analyst and a partner with intimate knowledge of the area.
- eCognition is a powerful tool. I suspect the PCA is doing the heavy lifting in the data fusion.
- The data fusion works best when the products are made from the same sensor. Other eCognition algorithms will fuse raster and vector data, but not the supervised classification algorithm.
- In the future, I recommend building your own indices outside of eCognition's collection of algorithms. Their NDWI uses the green band to highlight waterbodies but the green is also sensitive to built-up land like burn scars. This can conflict with NBR and NDBI.



The preservation of gallery forests is a critical need for savanna chimpanzee conservation. This project provided a reasonable roadmap for using satellite imagery and advanced GIS techniques to support these conservation efforts.



References and Resources

Boyer Ontl, K. M. (2017). Chimpanzees in the Island Of Gold: Impacts of artisanal small-scale gold mining on chimpanzees (*Pan troglodytes verus*) in Fongoli, Senegal (Doctoral dissertation). Ames, Iowa: Iowa State University.

Diallo, D., Ndiaye, P. I., Badji, L., & Pruetz, D. J. (2024). Savannah chimpanzee (*Pan troglodytes verus*) nesting behavior in the unprotected area of Tikankali near to a mining exploitation and the Niokolo Koba National Park in Senegal. *Frontiers in Ecology and Evolution*, 12(1228373).

Ndiaye, Y. H., Ndiaye, P. I., Lindshield, S. M., & Pruetz, J. D. (2024). Updating Chimpanzee Nesting Data at Mount Assirik (Niokolo Koba National Park, Senegal: Implications for Conservation. *Animals*, 14(553).

Gold Price Tracker:

<https://goldbroker.com/widget/historical/XAU?height=500¤cy=USD>

Google Earth Timelapse for Kedougou, Senegal Mining and Development from 1984 to 2022:

https://byclaudette.com/MGIS_assets/Timelapse.html

Mako Gold Mine: 12.85715,-12.3813

Bantakokouta Intermediate-scale Artisanal Mines: 12.76221,-12.22899

New Endeavor Mine: 12.964235, -12.098134

Sabodala Gold Mine: 13.197605,-12.114124

Copy and paste the coordinates into the search field, click Enter, set the speed to 1x and then press Play.

MODIS LST Tutorial:

MODIS Land Surface Temperature (LST) Annual Timeseries using Earth Engine by Spatial eLearning.

<https://youtu.be/0ASsr6Hj6NU?si=VHDrmLcHqpeRz1RV>

CHIRPS Precipitation Data Tutorial:

Download Climate Data (Rainfall) from 1981-2022 using Earth Engine API by Spatial eLearning.

<https://youtu.be/TcpG6SbUiYU?si=IFV7KzxpfbJn7hTX>



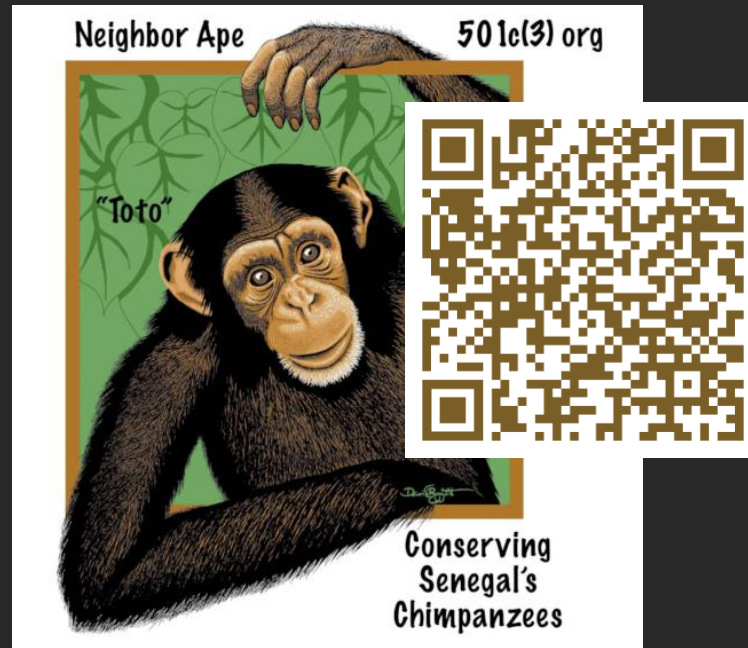
Questions?

Thank you!

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GEE Code and Links



Grassroots Funding

