

### **Remote Sensing Solutions to Assess Land Degradation and to German Remote**

# **Sensing Data Center** (DFD) Land Surface Dynamics

# **Enhance Smallholder Farming in West Africa**

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## BACKGROUND

### **Challenges in Sub-Saharan Africa**

- ... Sub-Saharan Africa faces **multiple risks (climatic,**
- social, economic, ecological) with partly
- uncertain future developments
- Projected **population development** from **400** million to 1.2 billion in 2100
- Different risks and different impact of land use in different agro-ecological zones and under farmers' socio-economic conditions
- Limiting yields and thus the acceptance of certain measures by farmers
- . This generally leads to agricultural expansion to satisfy increased demands and compensate degradation
- **Rising population** as driver for **increase in** cropland and in livestock



### **Research on Mitigation and Adaption**

in different projects

## CONCERT

Mitigation: • Greenhouse gas emissions and mitigation options are evaluated

![](_page_0_Picture_21.jpeg)

Suitability analysis for rice cultivation using climate and soil data Assessment of the actual used rice area using Sentinel-1 and Sentinel-2 data

#### Adaptation:

AgRAIN

PARTNERSHIPS FOR THE GOALS

8

- Co-Design with local farmers
- Identify socio-economic insentives for using SI practices
- Evaluation using EO data

![](_page_0_Picture_27.jpeg)

Consequences are loss in carbon and loss in biodiversity

Mapping actual rice cropping area in inland valleys

# METHODS

![](_page_0_Figure_30.jpeg)

![](_page_0_Figure_31.jpeg)

**Delineation of field boundaries using a CNN in Senegal** 

- **Input:** Planet data (NICFI) 2016-07 2023-07 (5 bands: R, G, B, NIR, maxNDVI)
- Use of maxNDVI to map all active fields
- Training data creation:
  - 4 sample regions representing different agricultural systems
  - Polygons were drawn across all agricultural fields
  - Rasterized to Planet resolution

#### Preprocess data:

Cropping/forest

Slope

Cacoa, cotton, mango etc.

Stretch the dataset using augmentation techniques like flipping, rotating and scaling and image generators

### Train the U-Net model (Ronneber et al., 2015):

- Tune hyperparameters and monitor learning curves to reach minimum loss
- fully convolutional neural network (CNN)
- Only convolutional layer

**Comparison** of field boundaries delineated from Planet (NICFI) data with:

- RGB high resolution image and ESA WorldCover
- Field data provided to local partner to plan with exact field sizes

Comparison of optimizer functions Adam optimizer Stochastic gradient descent

### Esa WorldCover & Field Boundaries

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## RESULTS

#### SPONSORED BY THE

![](_page_0_Picture_52.jpeg)

Federal Ministry of Education and Research

This work is part of the COINS project funded by the German Federal Ministry of Research and Education (Förderkennzeichen 01LL2204A-F)

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für Luft- und Raumfahrt

German Aerospace Center

EO-based analysis show that 2% of the country area is used for rice cropping

Ronneberer, O.; Fischer, P.; Brox, T. (2015): Medical Image Computing and Computer-Assisted Intervention (MICCAI), Springer, LNCS, Vol. 9351: 234-241.

Manobi Africa

Mapped areas follow inland valley 

**WASCAL** 

- North to south gradiant caused by the climate
- Rice cropping can increased from ~2-10% land area

![](_page_0_Figure_59.jpeg)

Mangroves

Moss & Lichen

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German Aerospace Center (DLR)

Earth Observation Center (EOC)

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Real-world" field boundaries from the AgCelerant platform provided by Manobi Africa

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