

AI4EO Challenges in the context of the Great Green Wall Initiative

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Great Green Wall Initiative

The GGW is a plan to build wall of trees across the African continent – from Senegal in the west to Djibouti in the east.

The GGW will act as a barrier to prevent spread of the desert.



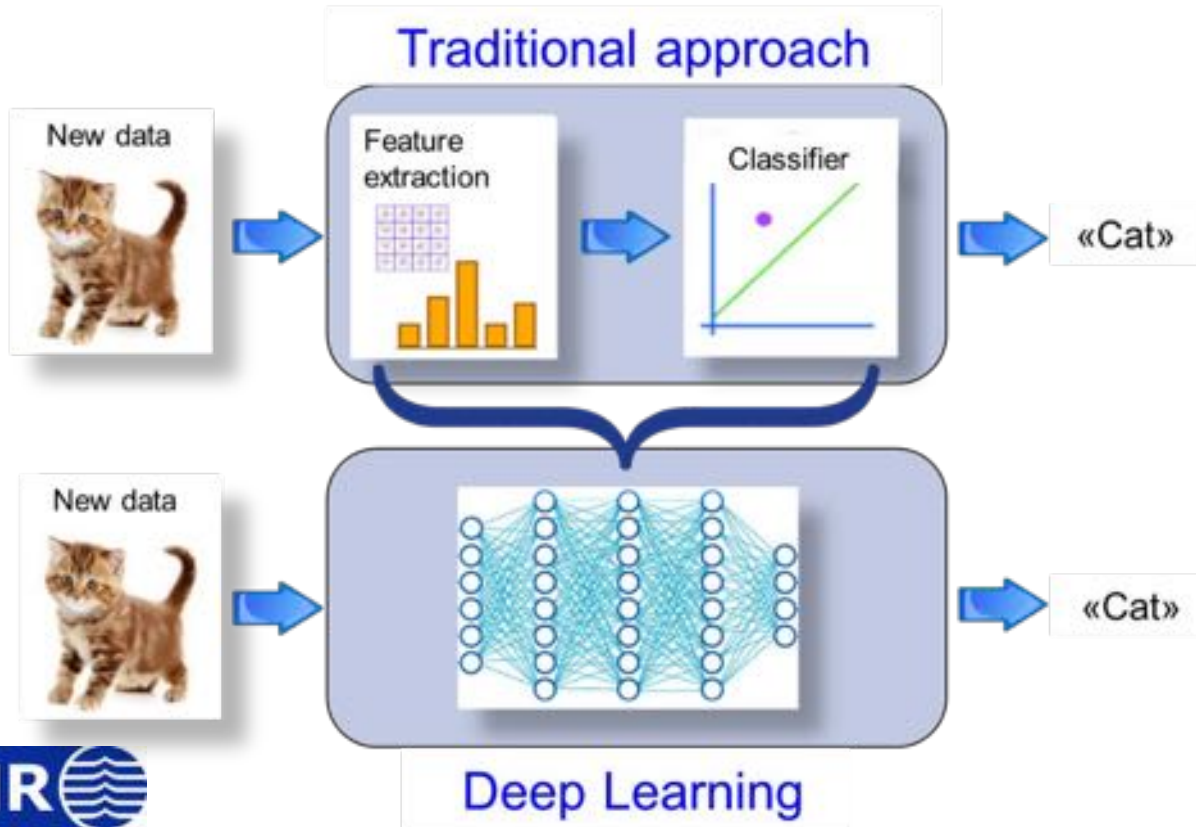
Great Green Wall Initiative

Goal

Explore if deep learning techniques applied to time series of Sentinel-2 data can detect and map dry forest in the Sahel regions



Deep Learning learns end-to-end



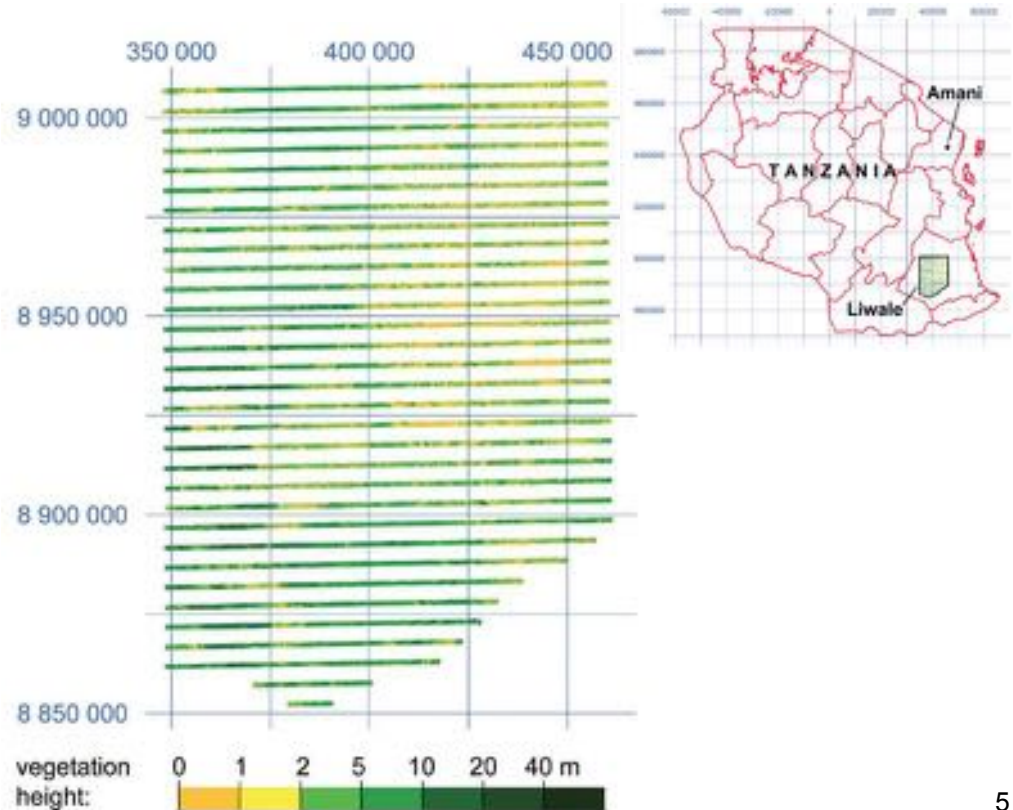
Realized using deep neural networks

Training data

We did not have annotated data available from the Sahel region

But we had lidar data from Liwale, Tanzania (2014)

From the lidar data we processed the **average tree height [m]** per Sentinel-2 pixel



Training data

Corresponding Sentinel-2 data:

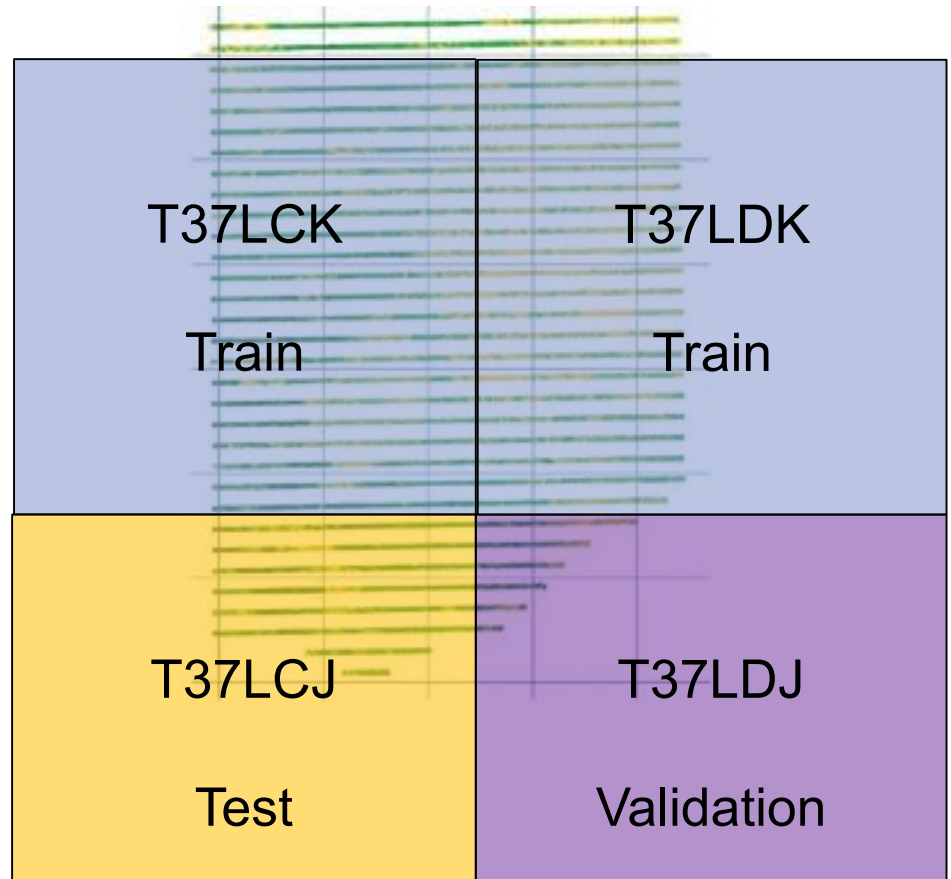
- Training tiles: T37LCK + T37LDK
- Validation tile: T37LDJ
- Test tile: T37LCJ

Test data from the Burkina Faso

- Tiles: T30PWT + T30PXT

Top of the atmosphere reflectance
and 10 bands (10m & 20m)

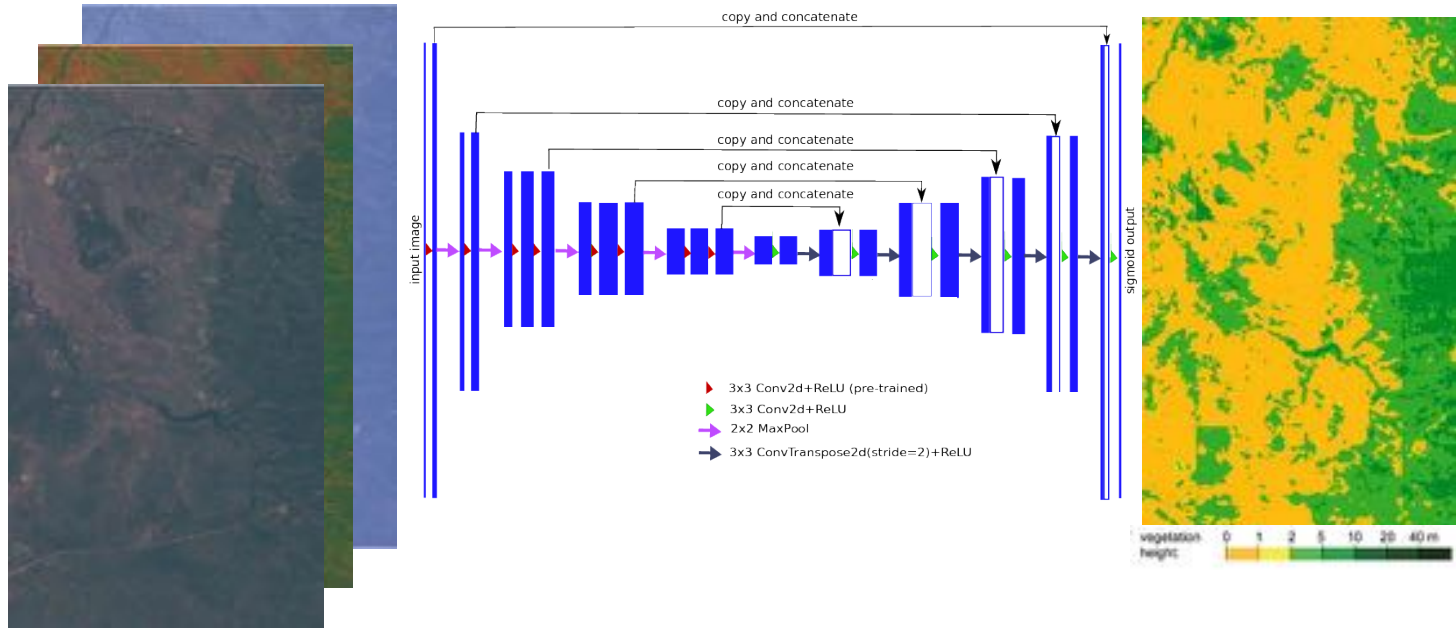
Cloud detection using S2Cloudless



Deep convolutional neural network

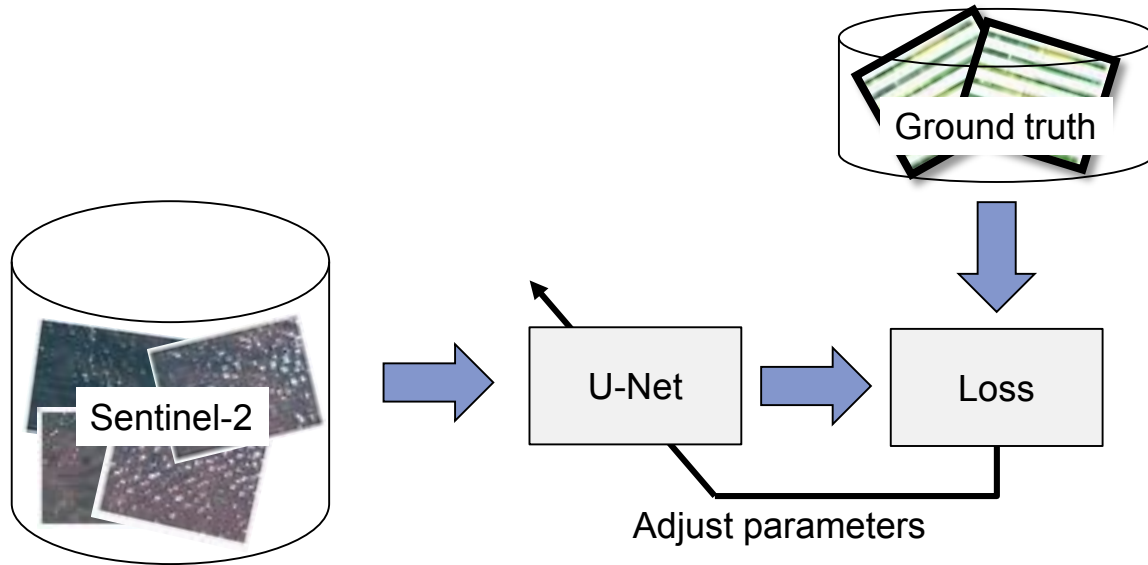
10 bands

Tree height or forest extent

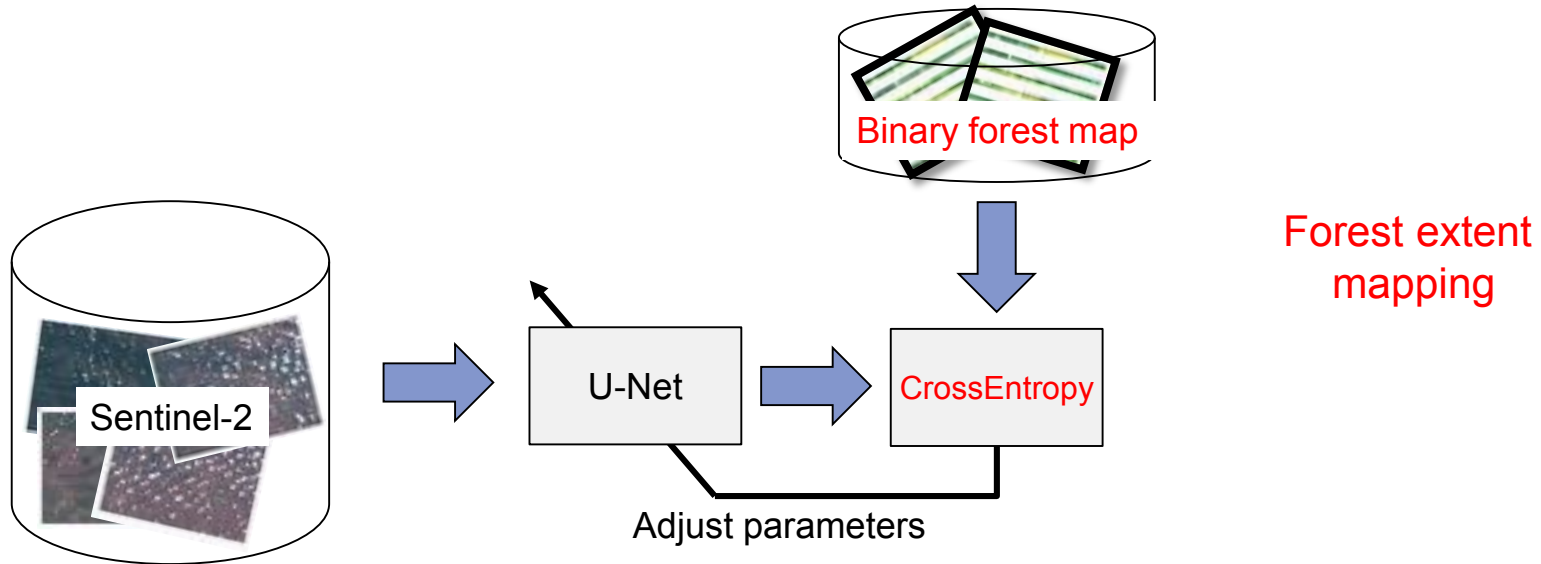


Network: U-Net (Ronneberger et al.)

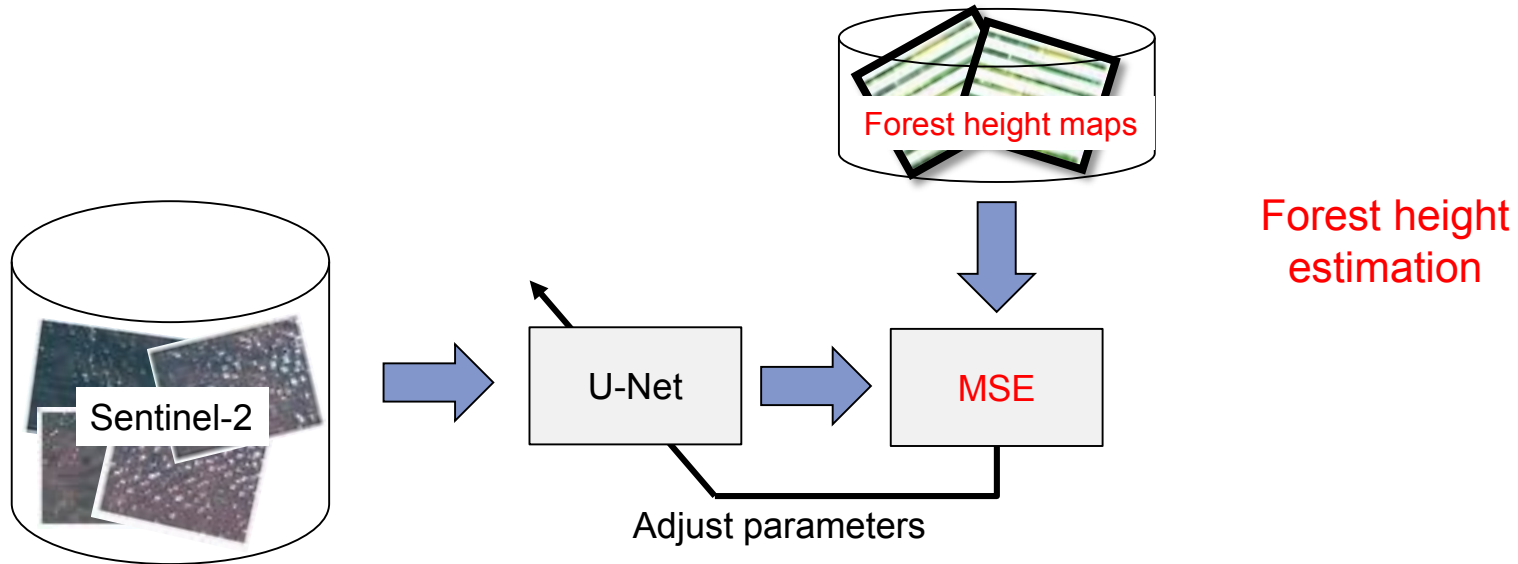
Training the deep neural network



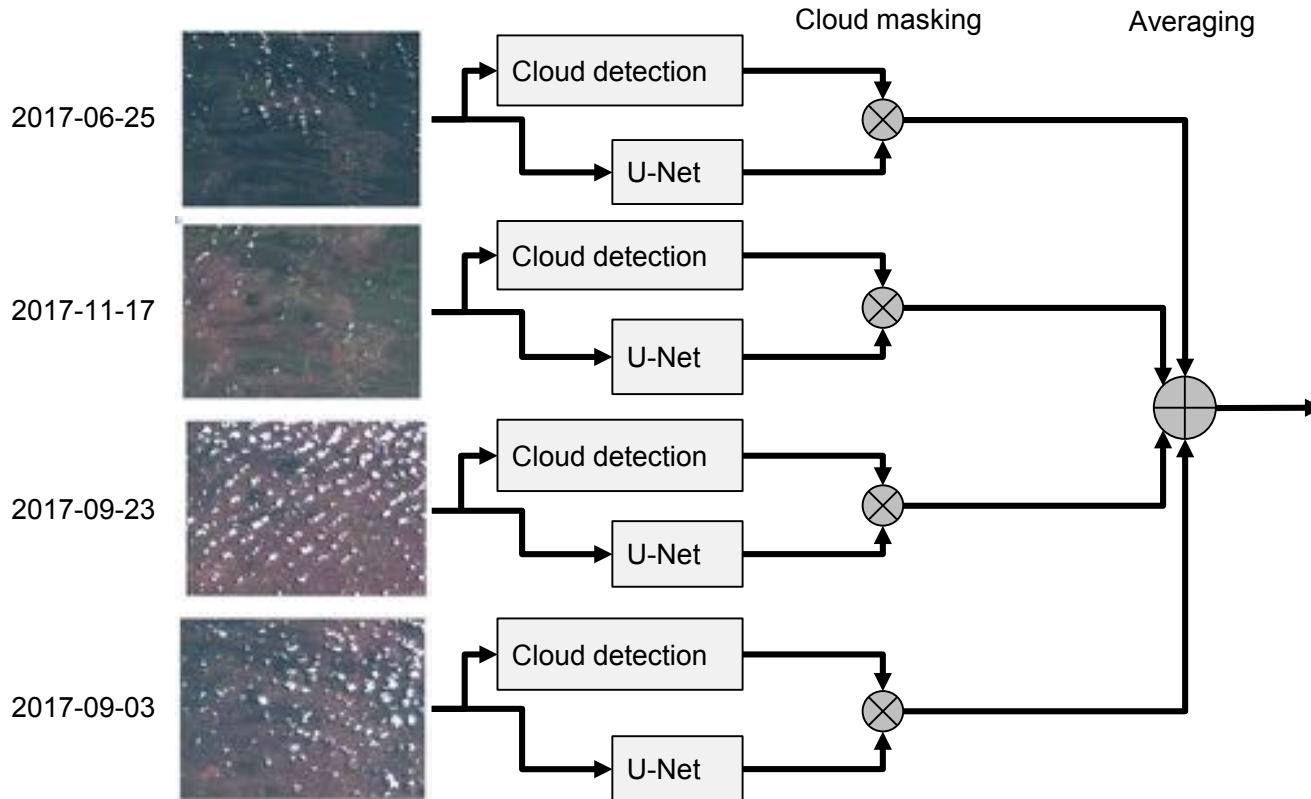
Training the deep neural network



Training the deep neural network



Merging multiple predictions



Results – Forest extent mapping, Liwale

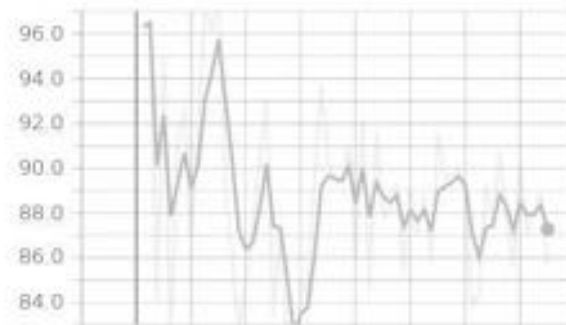
Mean tile NDVI > 0.6

Test tile: T37LCJ

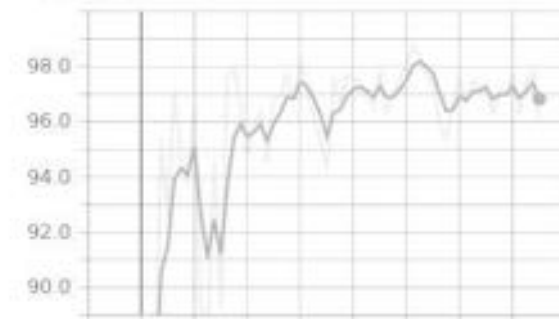
Configurations

- UNet
- Adam optimizer, LR=0.0004
- Median frequency balancing

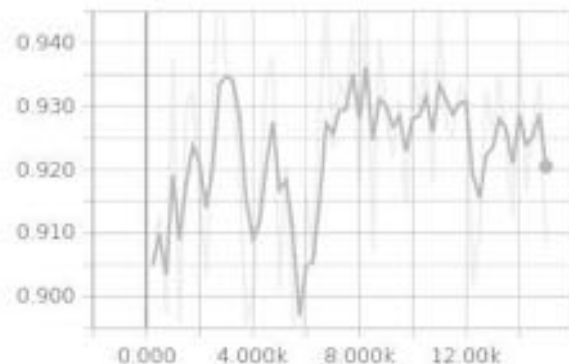
test_set_acc_class_0



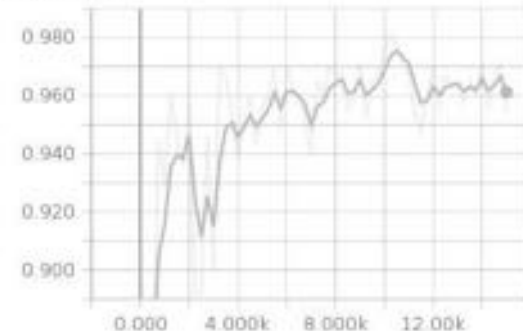
test_set_acc_class_1



test_set_acc_balanced



test_set_acc



Results – Forest extent mapping, Liwale

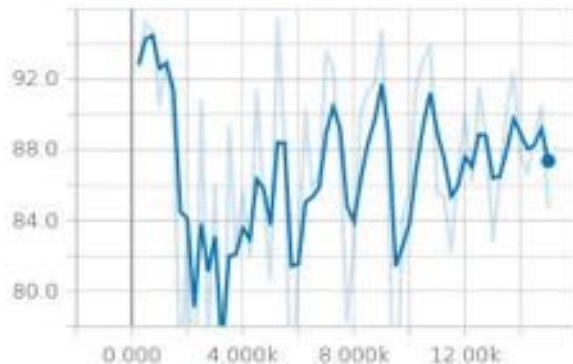
Mean tile NDVI: [0.4 – 0.5]

Test tile: T37LCJ

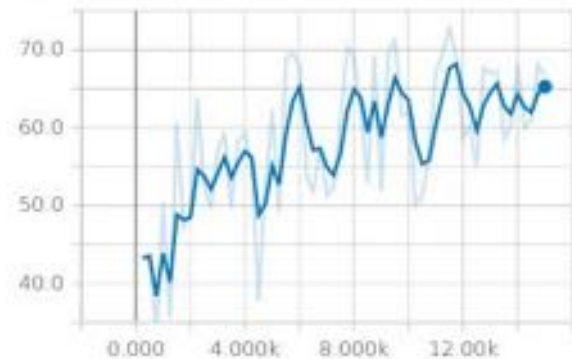
Configurations

- UNet
- Adam optimizer, LR=0.0004
- Median frequency balancing

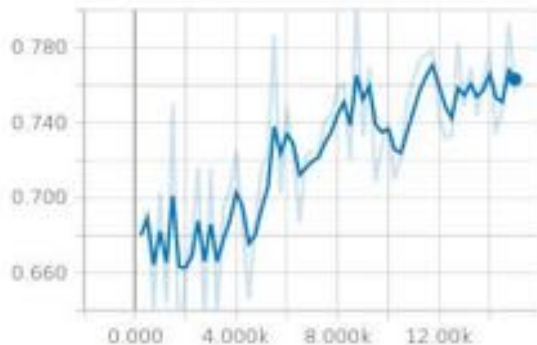
test_acc_class_0



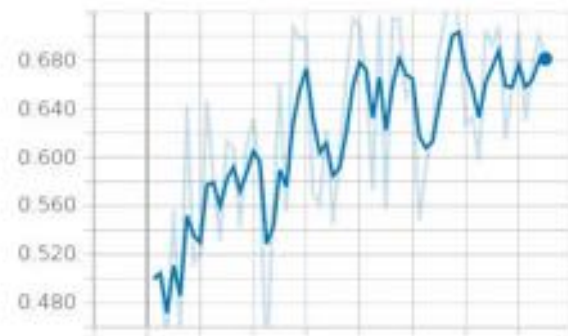
test_acc_class_1



test_acc_balanced

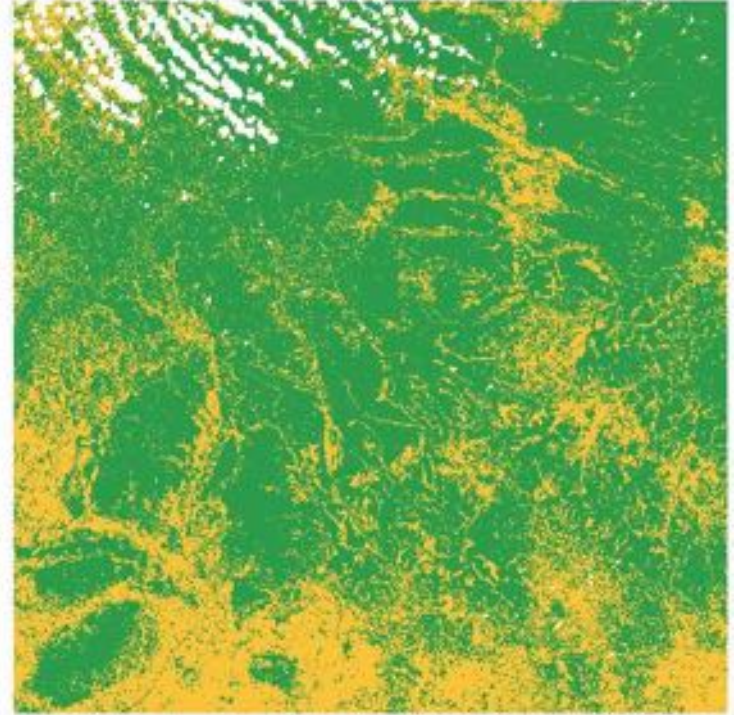


test_acc



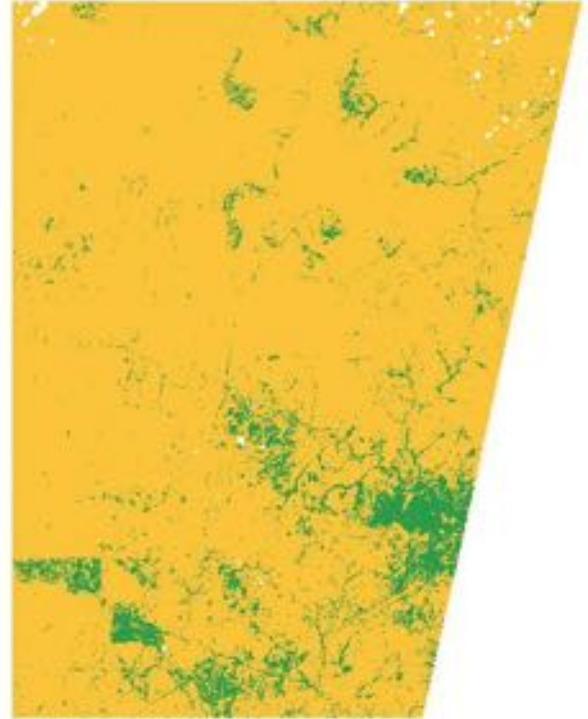
Results – Forest extent mapping, Liwale

Tile: T37LCJ



Results – Forest extent mapping, Sahel

Tile: T30PXT

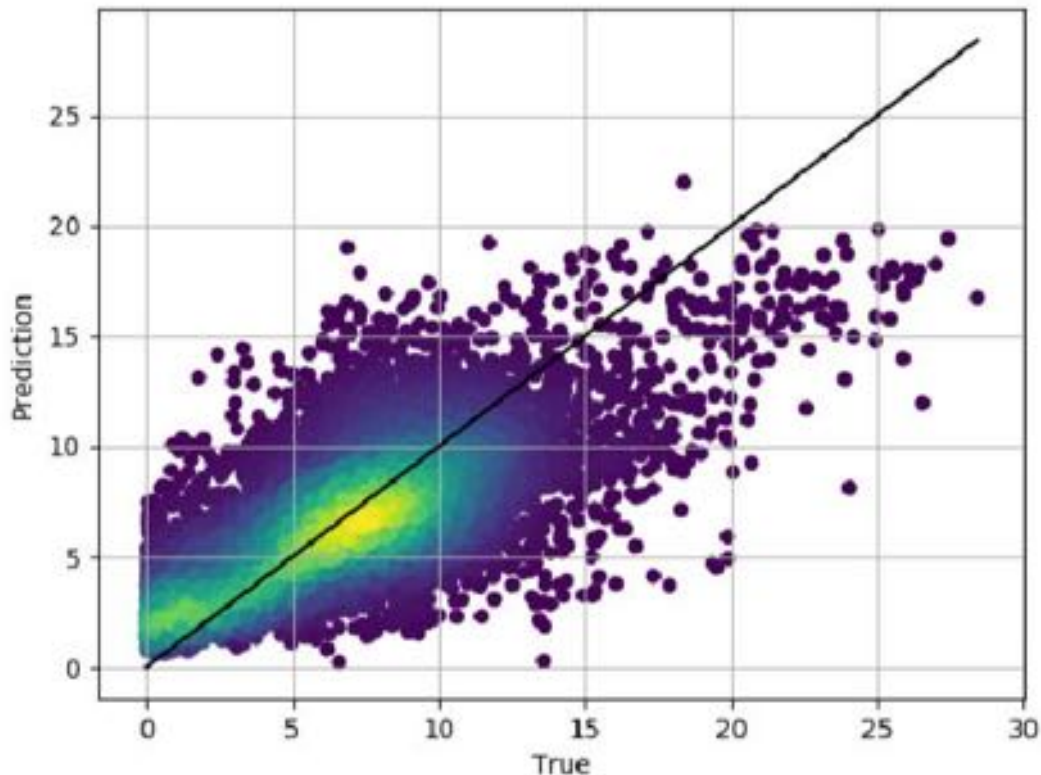


Results – Three height prediction, Liwale

Test tile: T37LCJ

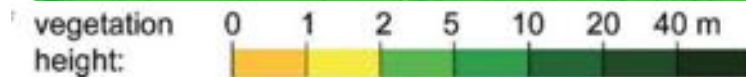
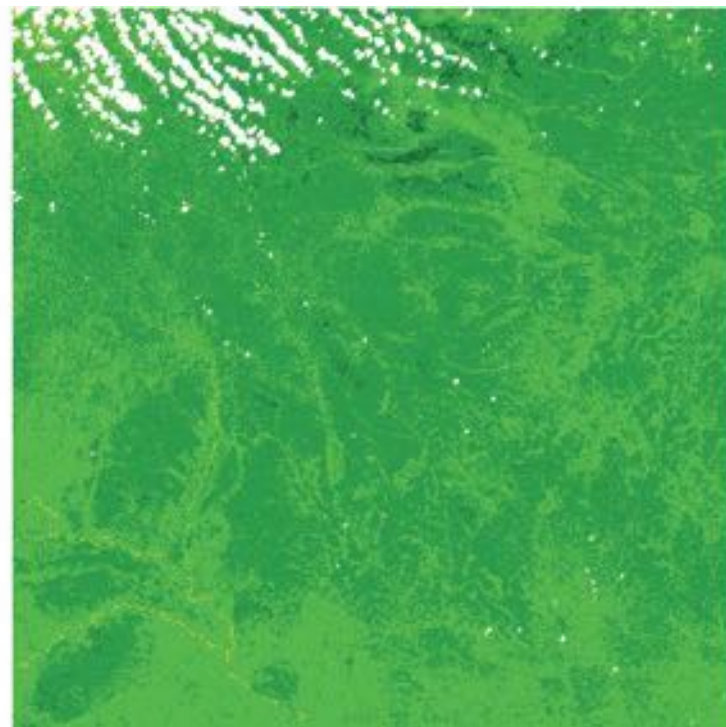
Configurations

- Mean tile NDVI > 0.6
- UNet
- Adam optimizer,
LR=0.0004
- Balanced histogram
sampling



Results – Three height prediction, Liwale

Tile: T37LCJ



Conclusions

- ▶ The deep learning based valid chain is flexible. The same chain is used for both forest extent mapping and tree height estimation.
- ▶ The results depends strongly on the greenness of the vegetation. Tiles with high mean NDVI values gave better results than tiles with low NDVI values.
- ▶ The results of the tree height prediction is in general good, but the algorithm struggles to predict tree heights above 15m
- ▶ The results for Burkina Faso (Sahel) need to be evaluated properly.
- ▶ Further studies will be to investigate multi-sensor approaches (include Setinel-1) and surface reflectance data