Using deep learning and remote sensing to map and track land use following deforestation across Africa

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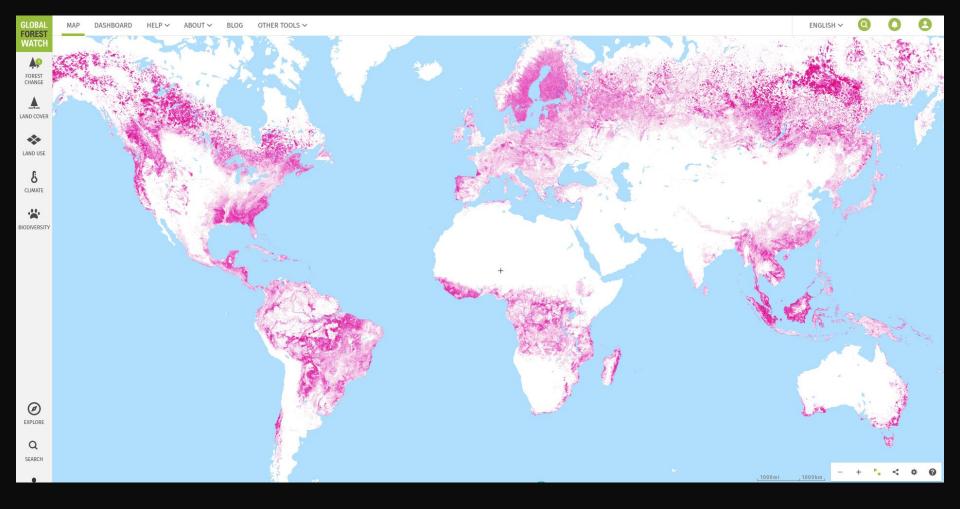












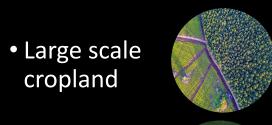
Motivation

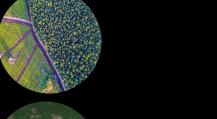
Knowing what causes this tree cover loss can help tackle deforestation

The underlying causes may be very complex, but we can at least see what happens with the land once it is deforested:

The Follow-up land use (FLU)

Land activities that follow deforestation

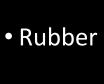
















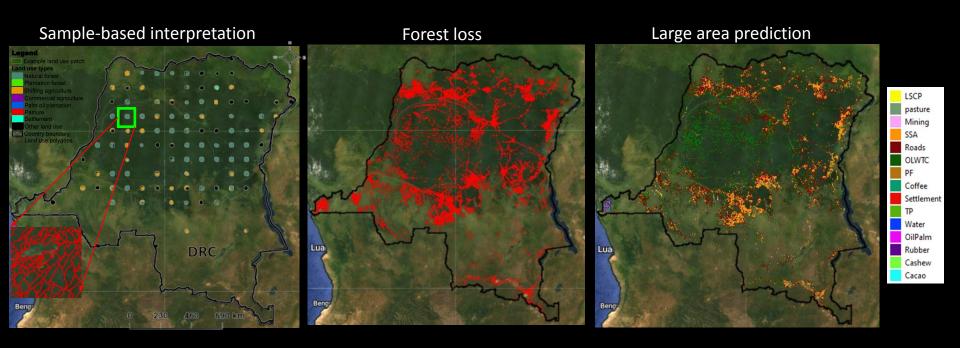








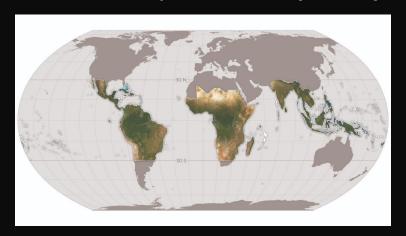
Need for automated large scale land use monitoring



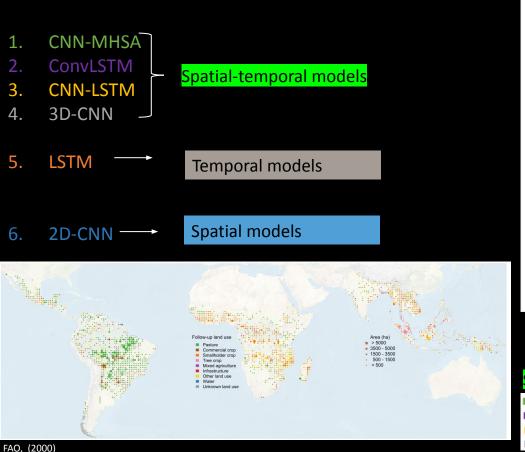
Freely available data sources

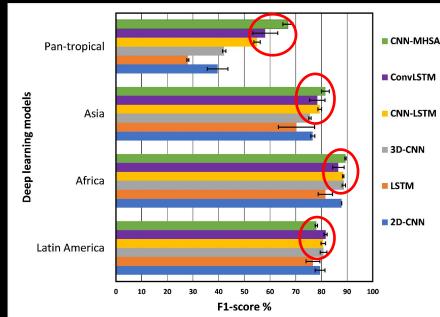
- Landsat
 - 30 m resolution
 - Multispectral (12 bands)
 - Revisit time of 16 days
- Sentinel 2
 - 10 m resolution
 - Multispectral (10 bands)
 - Revisit time of 5 days

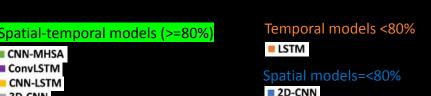
- Planet
 - 5 m resolution
 - RGB+NIR (4 bands)
 - Revisit time of 1 day
 - Norway pays for the tropical data! (NICFI)



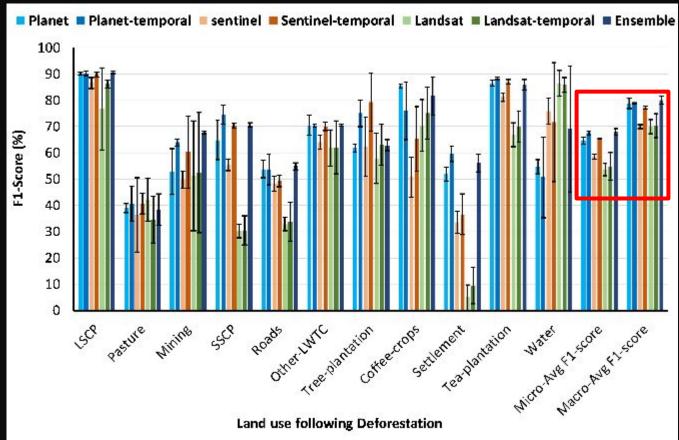
What deep learning architecture to use for assessing land-use following deforestation using remote sensing data?

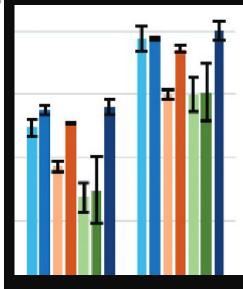




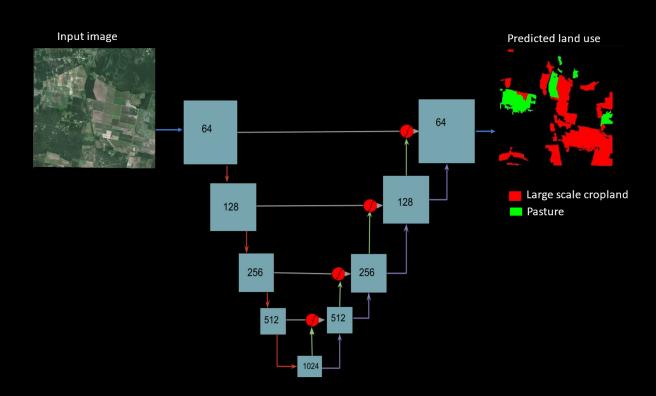


And what data source?





Model: Attention U-Net



Labeled data sources

Cashew

Х

X

X

X

X

Х

X

X

| Labeled data sources | | | | | | | | | | | | | |
|--|------|---------|--------|------|-------|-------|----|--------|------------|----|-------|---------|--------|
| Data | LSCP | Pasture | Mining | SSCP | Roads | OLWTC | PF | Coffee | Settlement | TP | Water | OilPalm | Rubber |
| FAO 2010 global Remote Sensing Survey | x | x | x | x | | x | | | X | | x | | |
| Crowdsourced deforestation drivers (IIASA) (Bayasa et al., 2022) http://pure.iiasa.ac.at/id/eprint/17539/) | x | х | х | x | x | | x | | x | | | x | |
| Masolele et al,. 2022 (Ethiopia) | x | x | x | x | x | x | x | x | x | x | × | | |
| ICRAF, Econometric | x | | | x | | x | | | x | | x | x | x |
| NAFORMA (Tanzania) | x | | x | x | | | x | | | | | | |
| Large-scale farms and small holder (Jann et al,. 2018) (Zambia) | x | | | x | | | | | | | | | |
| Global Map of Oil Palm Plantations (Descale et al., 2021) | | | | | | | | | | | | x | |
| Kenya GIS data (World Resources Institute | x | | | х | | | X | x | | x | | | |
| - https://www.wri.org/data/kenya-gis-data) | | | | | | | | | | | | | |
| Namibia | X | | | X | | | | | | | | | |
| Ghana | x | | x | x | | | | | | | | x | x |
| | | | | | | | | | | | | | |

X

X

X

Х

X

Google research open-buildings dataset

ngs/)

open-data/ (Mining)
Landuse data Nigeria

(https://grid3.gov.ng/datasets)

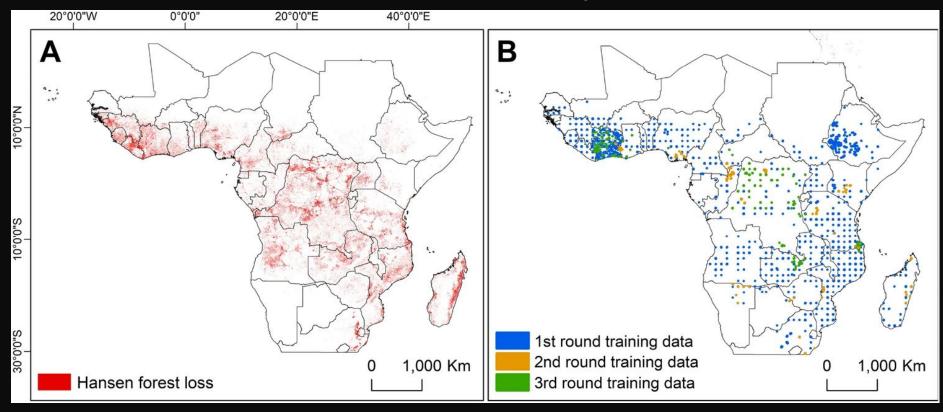
(https://sites.research.google/open-buildi

https://ipisresearch.be/home/maps-data/

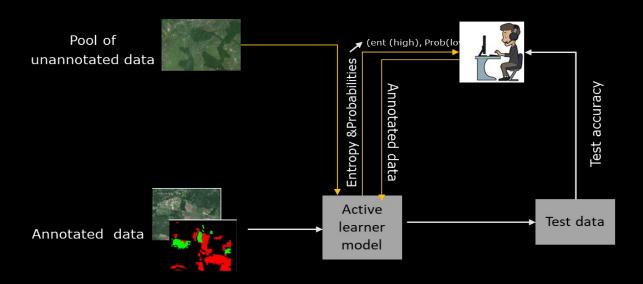
Results

Well, not good

The obtained labels were quite biased

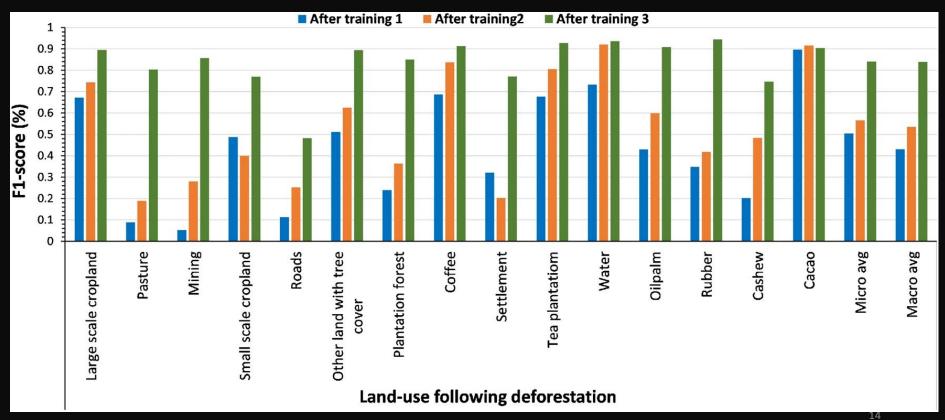


Active learning



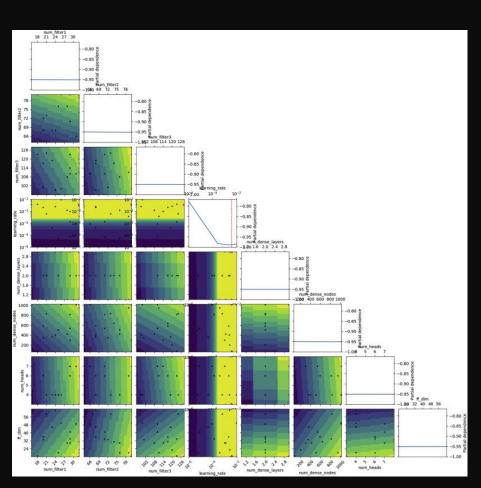
Improvement of accuracies with active learning

Active learning



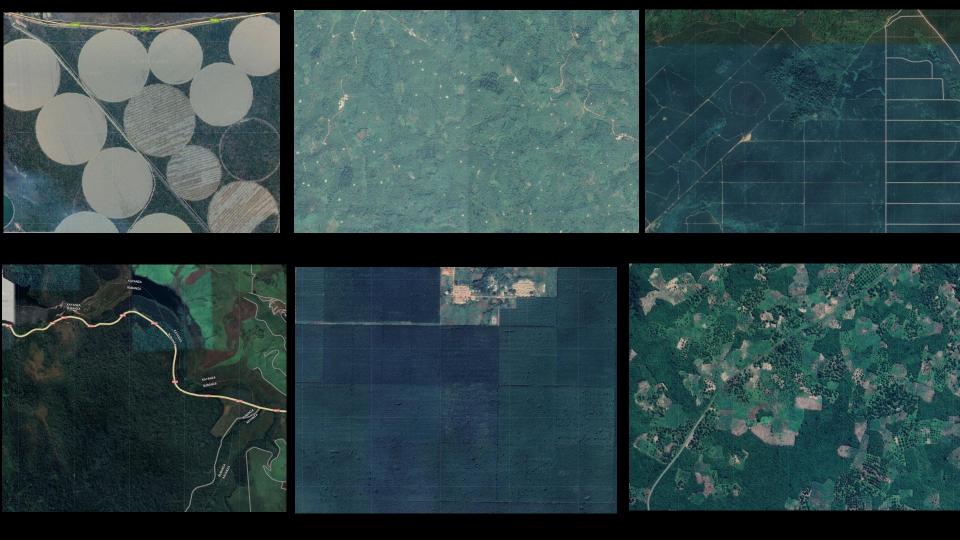
Hyperparameter optimization

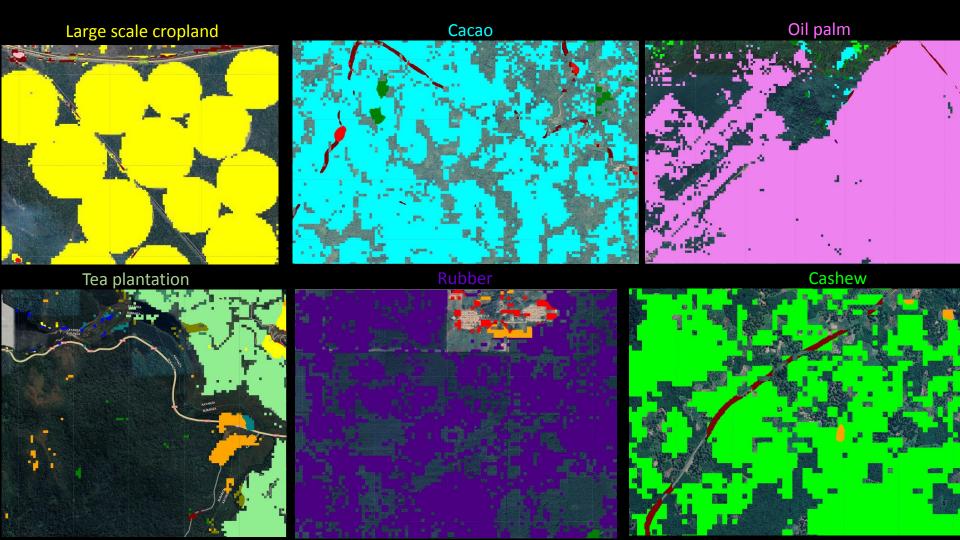
Partial Dependence plot, showing a matrix-plot of all combinations of searched hyperparameters via Bayesian optimization.



Results

Now yes!

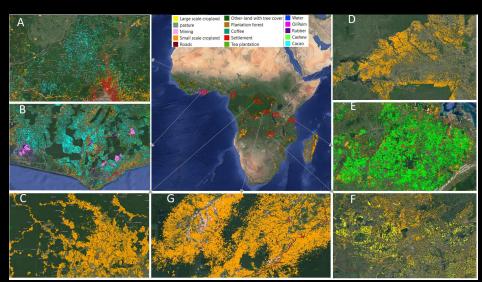




Results: Monitoring direct drivers of deforestation

- Annual follow-up land use 2001 2020 mapped at Hansen forest loss data (wall-to-wall for entire Africa)
- Planet NICFI data and deep learning approach used for mapping
- 15 land use classes
- High accuracies (82% average macro-F1, see next slides for more detail)





Masolele et al., under review in NSR

https://robertnag82.us ers.earthengine.app/vi ew/africalu

Results: Hotspot of land use following deforestation (Africa)

