



Sentinel-2 image of Cambridge  
Taken October 17, 2022

AI-Assisted Conservation Workshop

## AI for the Digital Planet(s)

Dr. Hannah Kerner

Assistant Professor, ASU School of Computing and Augmented Intelligence

AI/ML Lead, NASA Harvest





# Vision: AI for the Digital Planet



Ocean observations



Webcams and smartphones

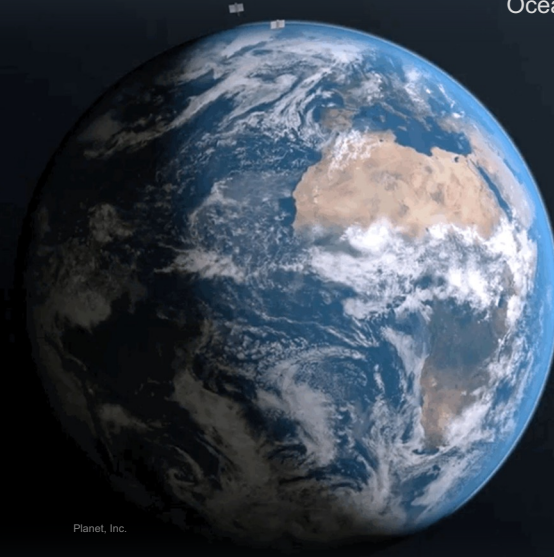
In-situ sensors



Proximal sensors



Suggs/NCSU

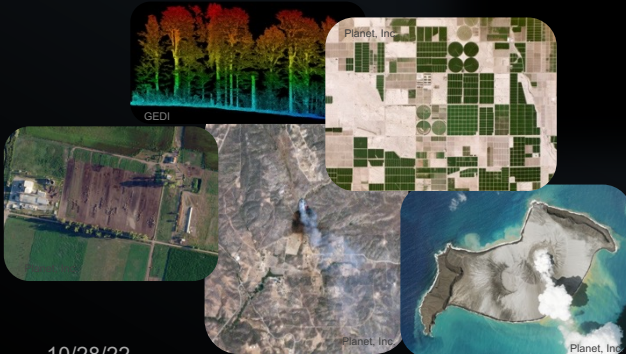


Planet, Inc.

Street-level images/videos



Satellite/spaceborne observations



## Use case / stakeholder need



Need census of  
smallholder  
farms in Togo

**Cina Lawson**

*Togolese Minister of Post,  
Digital Economy  
and Technological  
Innovation*

Use case /  
stakeholder need

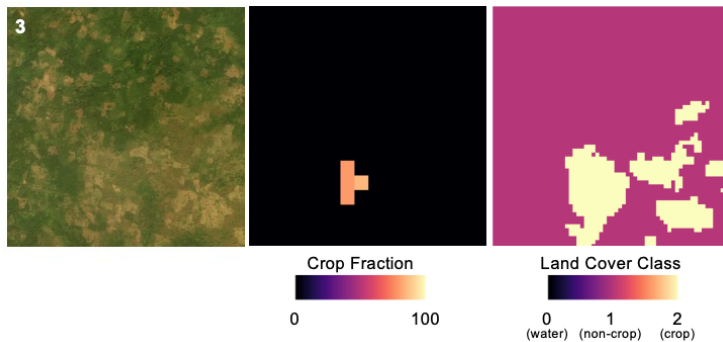


Existing AI methods  
insufficient?



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## Use case / stakeholder need



## Existing AI methods insufficient?

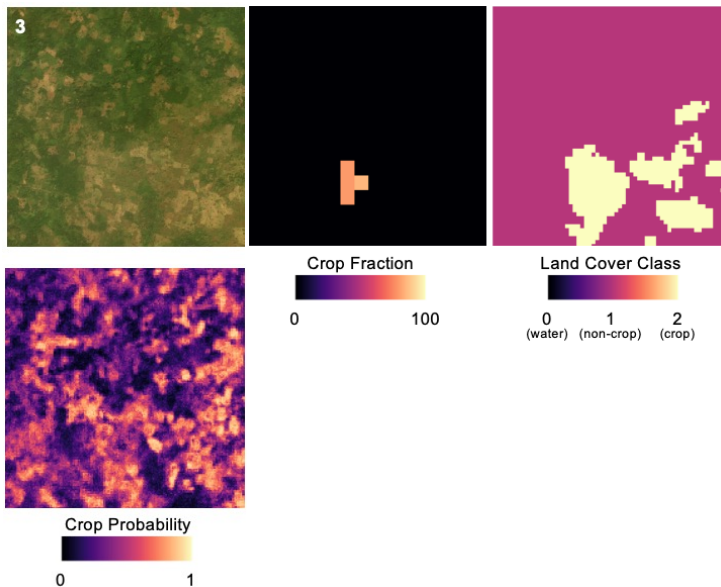


## Solution



Need census of smallholder farms in Togo

**CINA Lawson**  
*Togolese Minister of Post, Digital Economy and Technological Innovation*



“This map provides unmatched clarity into the nature and distribution of agricultural land nationwide [and helps] provide decisive knowledge being used to design social protection policies aimed at improving the livelihoods of agrarian rural communities.”

Deliver



Case /  
stakeholder need

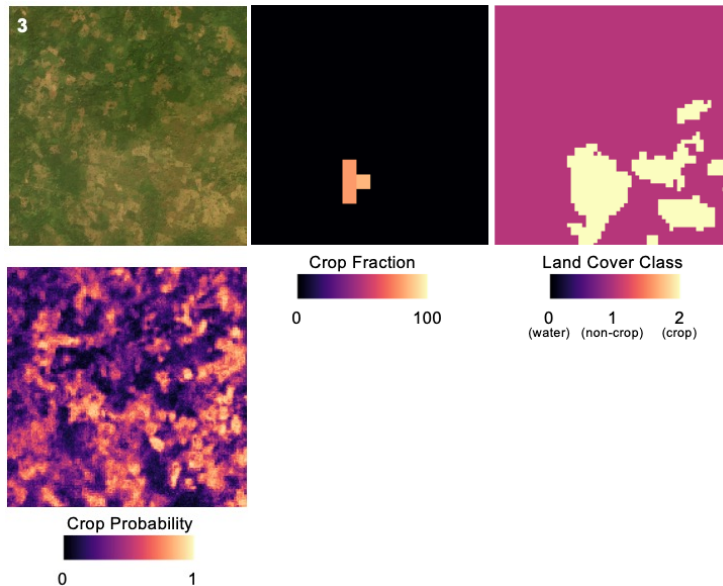
Existing AI methods  
insufficient?

Solution



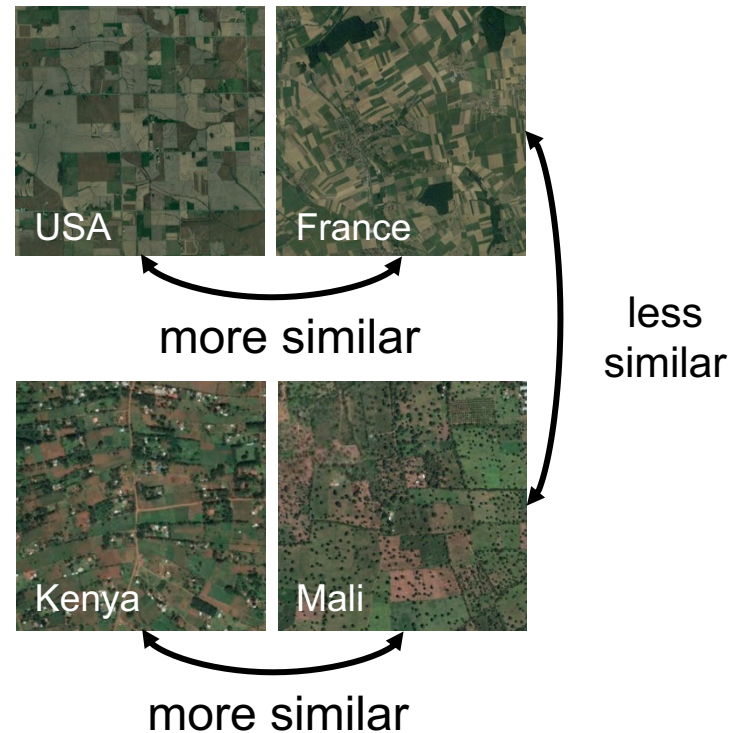
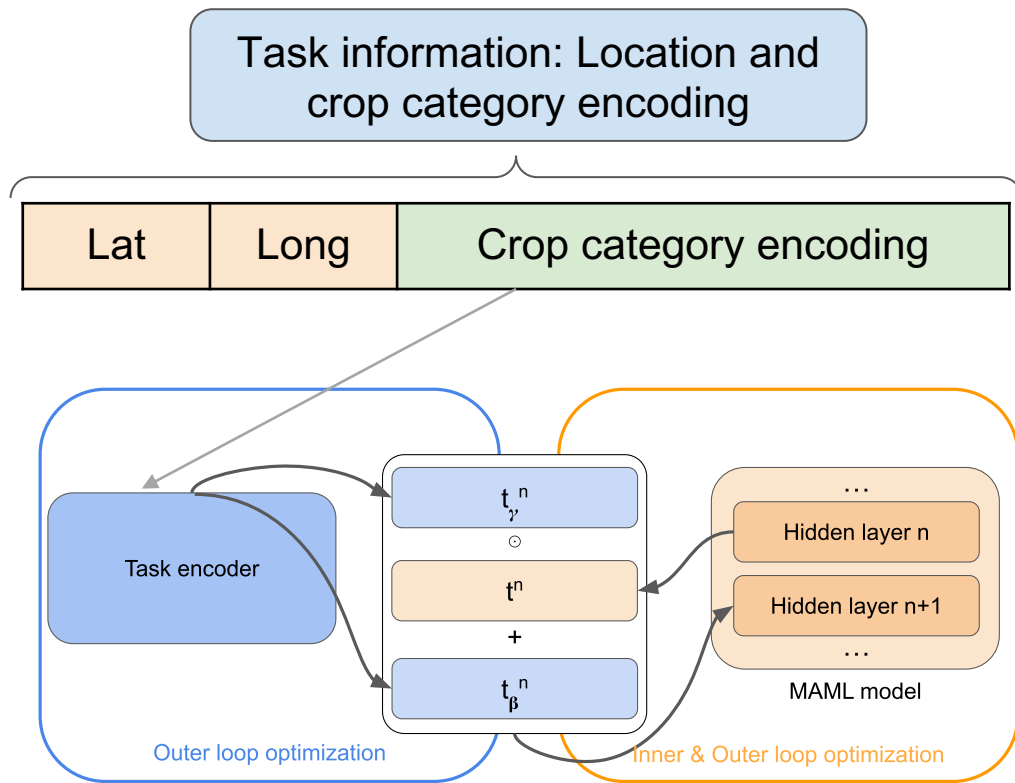
Need census of  
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# Learning from limited labels: Task-informed meta-learning



# Scalable geospatial prediction: OpenMapFlow

Rapid map creation with machine learning and  
earth observation data

**openmapflow 0.2.0rc1**

```
pip install openmapflow==0.2.0rc1
```



Data processing  
pipeline



ML model  
training



Rapid map  
creation

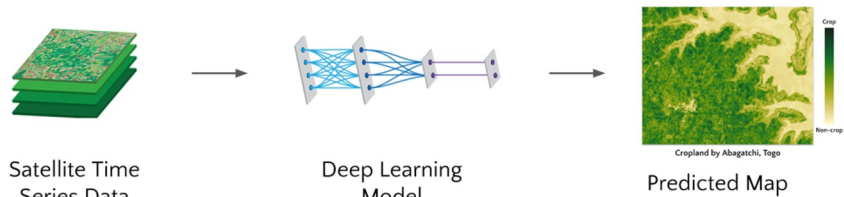
Reduces time and cost





# Scalable geospatial prediction: OpenMapFlow

## crop-mask-example 🌍



## buildings-example 🌍

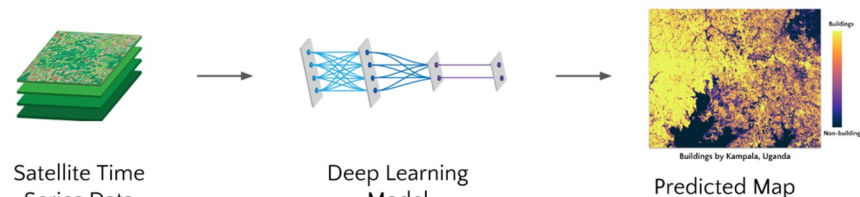
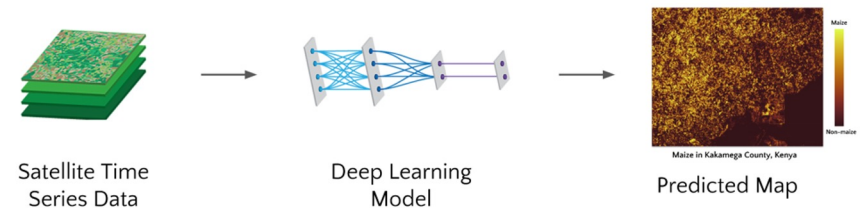


Table 1: Dense Predictions Time and Cost

Bounding Box	Area	Input data	Sequential prediction time	Our prediction time	Merging time	Our cost
Rwanda	63,018 km <sup>2</sup>	0.331 TB	458 mins	12 mins	7 mins	\$9.19
Western Ethiopia	160,036 km <sup>2</sup>	1.38 TB	1908 mins	19 mins	40 mins	\$42.50
Uganda	375,755 km <sup>2</sup>	3.06 TB	4232 mins	39 mins	101 mins	\$93.25
	1000 km <sup>2</sup>	8.83 GB	12 mins	1 min	<1 min	\$0.30

## maize-example 🌍

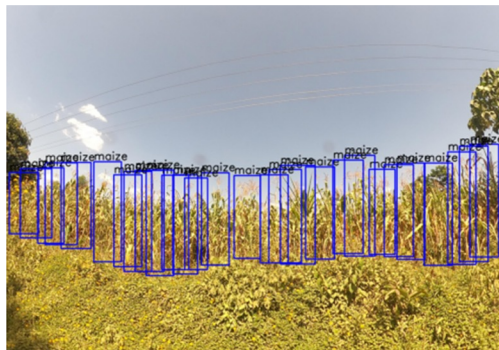


Zvonkov, I., Tseng, G., Nakalembe, C., Kerner, H. Under review for AAAI 2023.

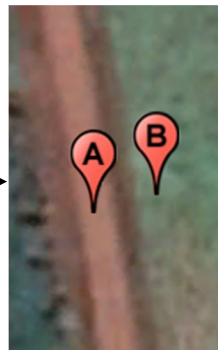
# Exponential data collection: Street2Sat



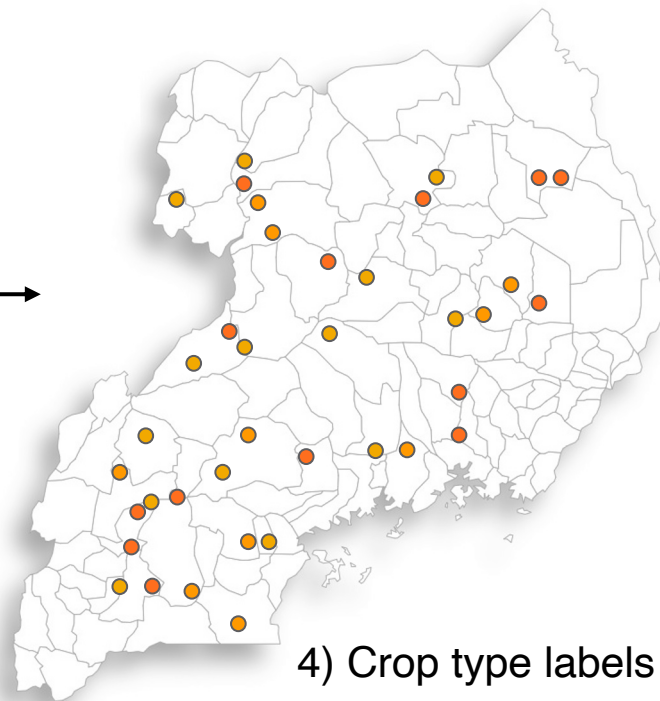
1) GoPro photos  
+ GPS coordinate



2) Crop type object  
detection



3) Location  
offset



4) Crop type labels

# Incorporating human expertise: Corrective labeling

## Corrective Labeling App - Malawi

This app allows users to add points to where the crop map is wrongly classified. Users can compare the cropland map on the left against the high resolution Sentinel 2 data on the right. The "Add a Marker" icon on the left panel can be selected to start creating the points. To delete a point, click the hand icon to select the specific point and hit the delete button. After picking all the points, users can click the "Get link for downloading points" button to generate a link that can be downloaded as a CSV file. A detailed instruction manual of the App can be found in the link below:

[Detailed Instructions](#)



[Get link for downloading points](#)



# Capacity building boosting

