

# From Forest Monitoring to Cultural Heritage: Reanalyzing LiDAR Data with Deep Learning to Map Ancient Structures in Yucatan, Mexico

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# Objective

Investigate the applicability of deep learning models, particularly Mask R-CNN, for detecting and segmenting archaeological structures based on airborne LiDAR data in the Puuc region of Yucatan, Mexico.





# Studiegebiet, Puuc-regio, Yucatan, Mexico

Huntichmul



Legende

0 0,5 1 2 km

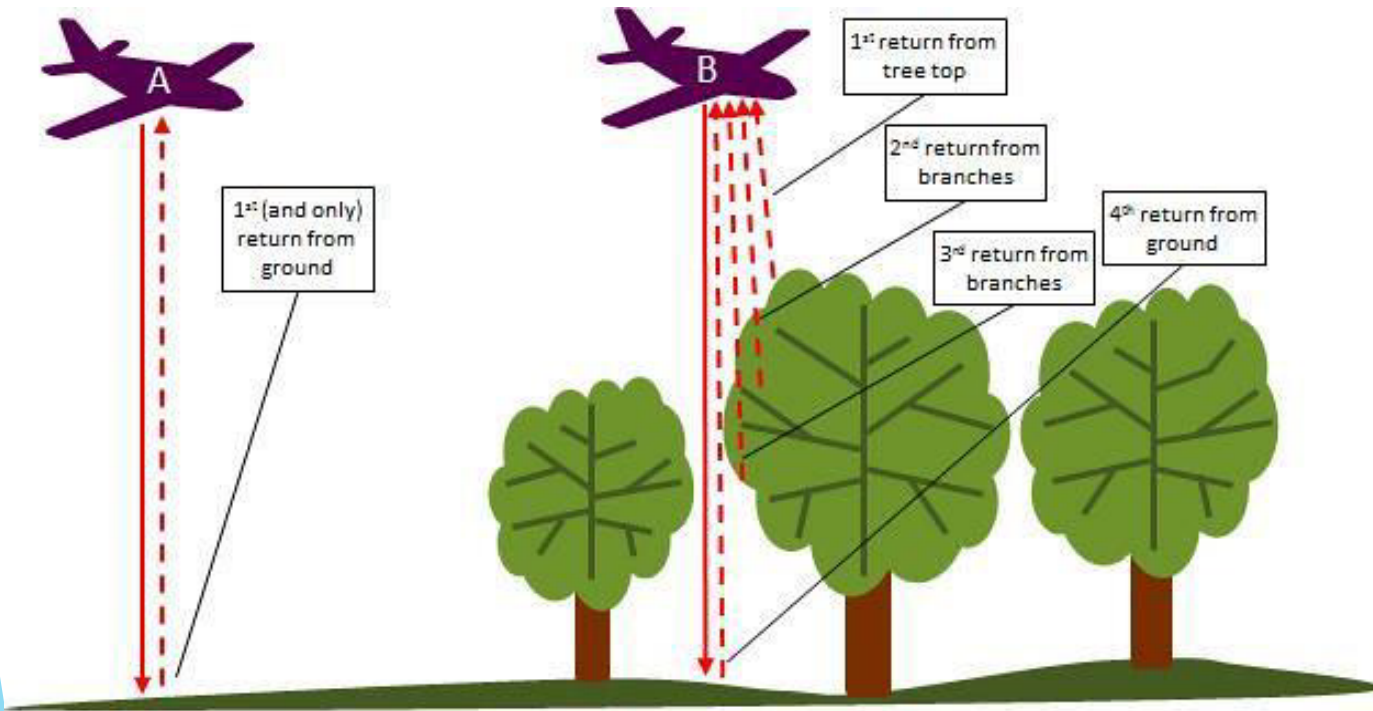


0 115 230 460 km

Guatemala  
HONDURAS  
San Salvador  
Tegucigalpa  
NICARA



# LiDAR data

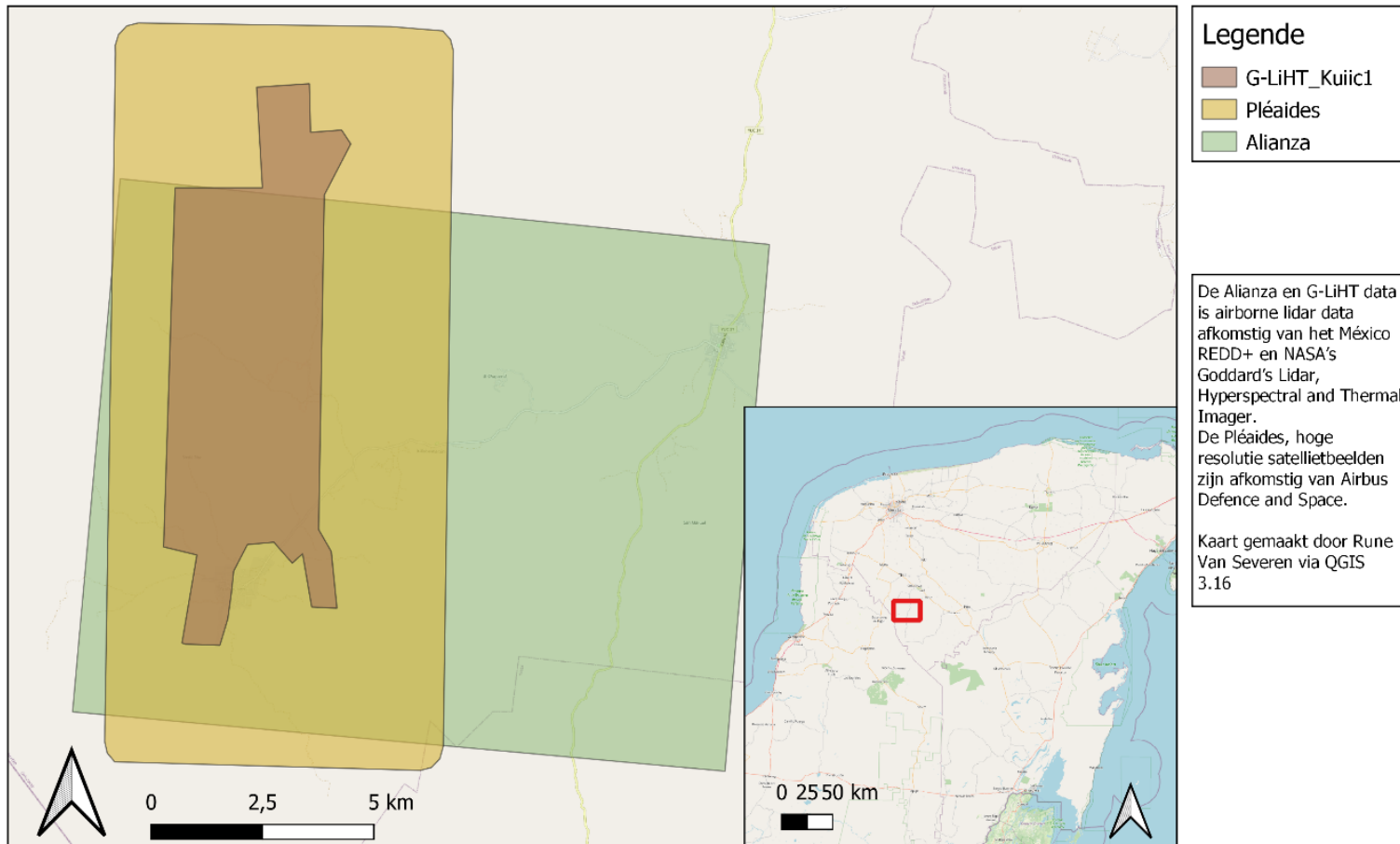


## LIDAR

- ❖ Active Sensor
- ❖ Provides 3D point clouds
- ❖ Detailed 3D information (DTM/DSM/CHM)
- ❖ Penetrates forest canopies



# LiDAR data



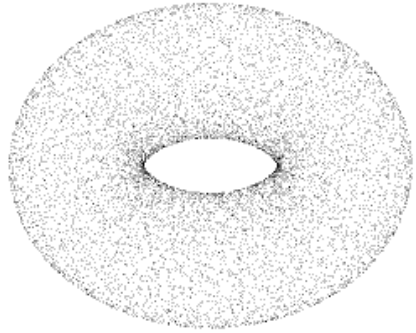
## LIDAR

- ❖ AlianzaRedd+ LiDAR data
- ❖ Woodwell Climate Center
- ❖ Funded by USAID (Cartodata)
- ❖ 29/03/14 - 22/4/14
- ❖ 175km<sup>2</sup> - 8p/m<sup>2</sup> - >100Gb

## OTHER DATA

- ❖ G-LiHT Data (NASA)
  - ❖ April/May 2013
- ❖ Pléiades Tri-stereo data
  - ❖ 15/01/2022

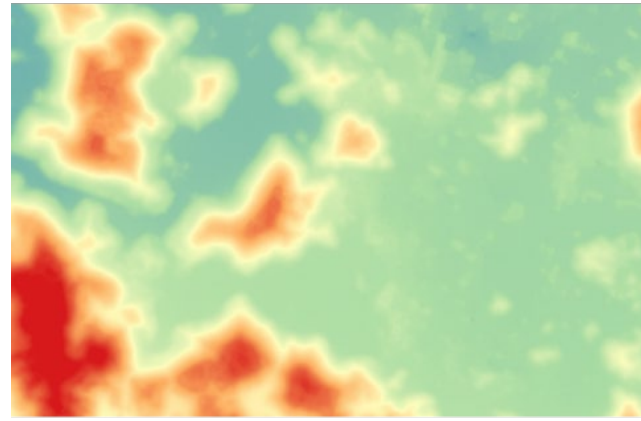
# METHODS



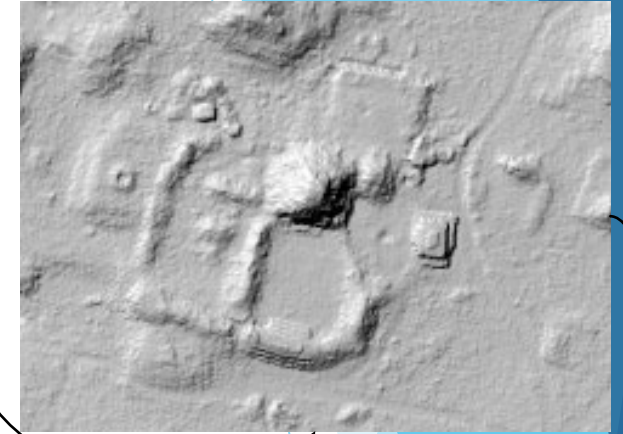
LIDAR POINT CLOUD

PREPROCESSING  
CLASSIFICATION  
GROUND/NON-GROUND

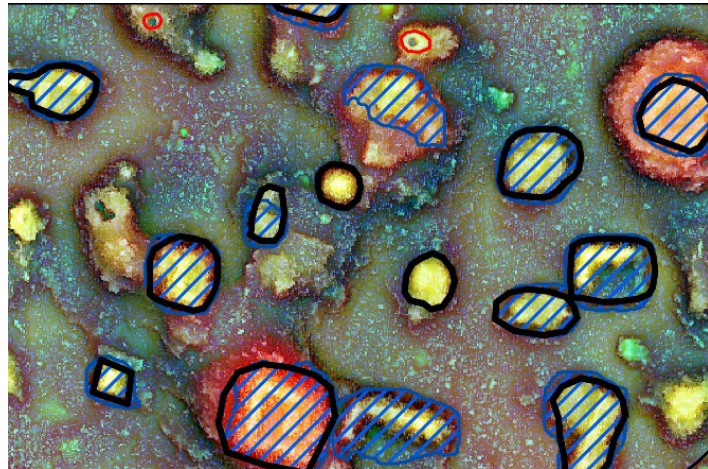
GENERATE DTM



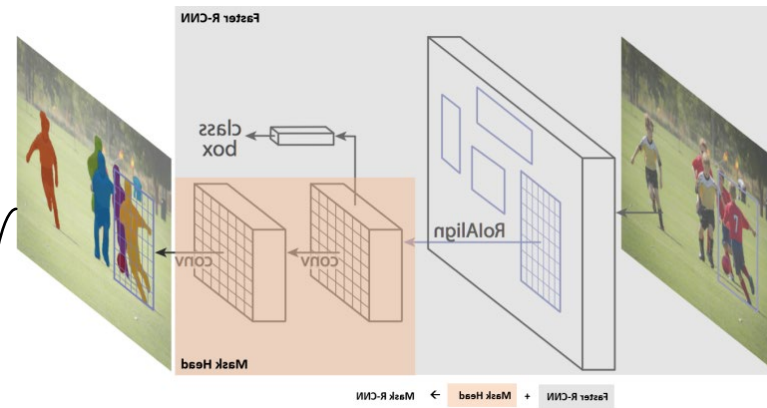
TERRAIN  
VISUALISATIONS



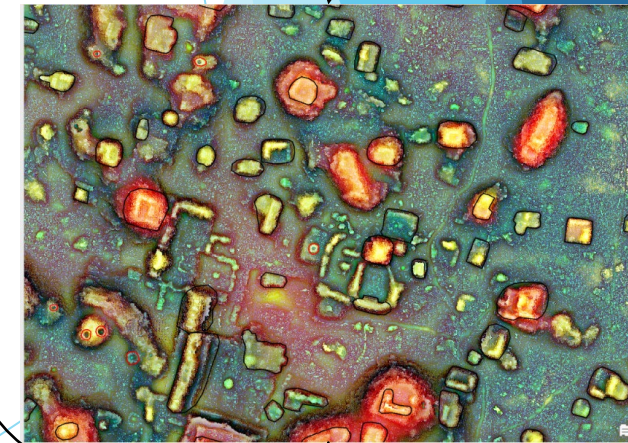
ACCURACY ASSESSMENT



OBJECT INSTANCE SEGMENTATION



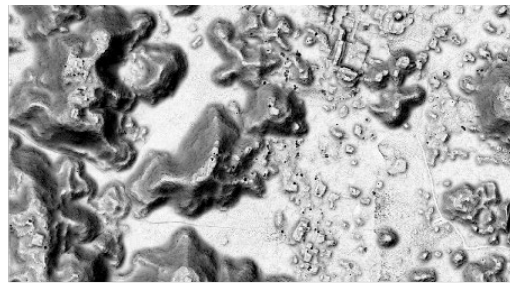
MASK-R CNN



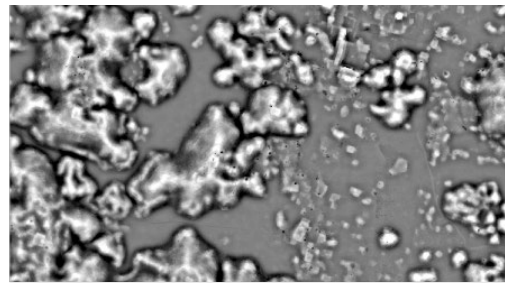
VISUAL SAMPLING



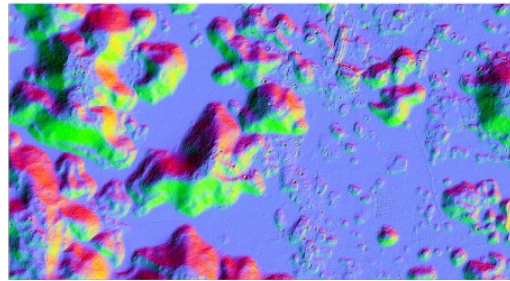
# Enhanced v3 Multi-scale topographic position (e3MSTP)



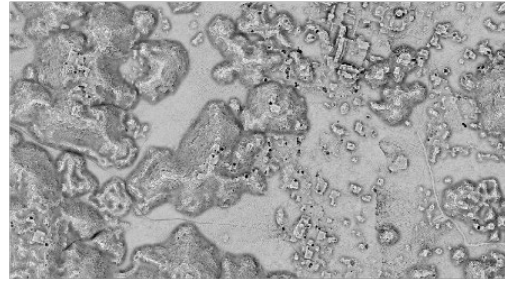
Sky view factor



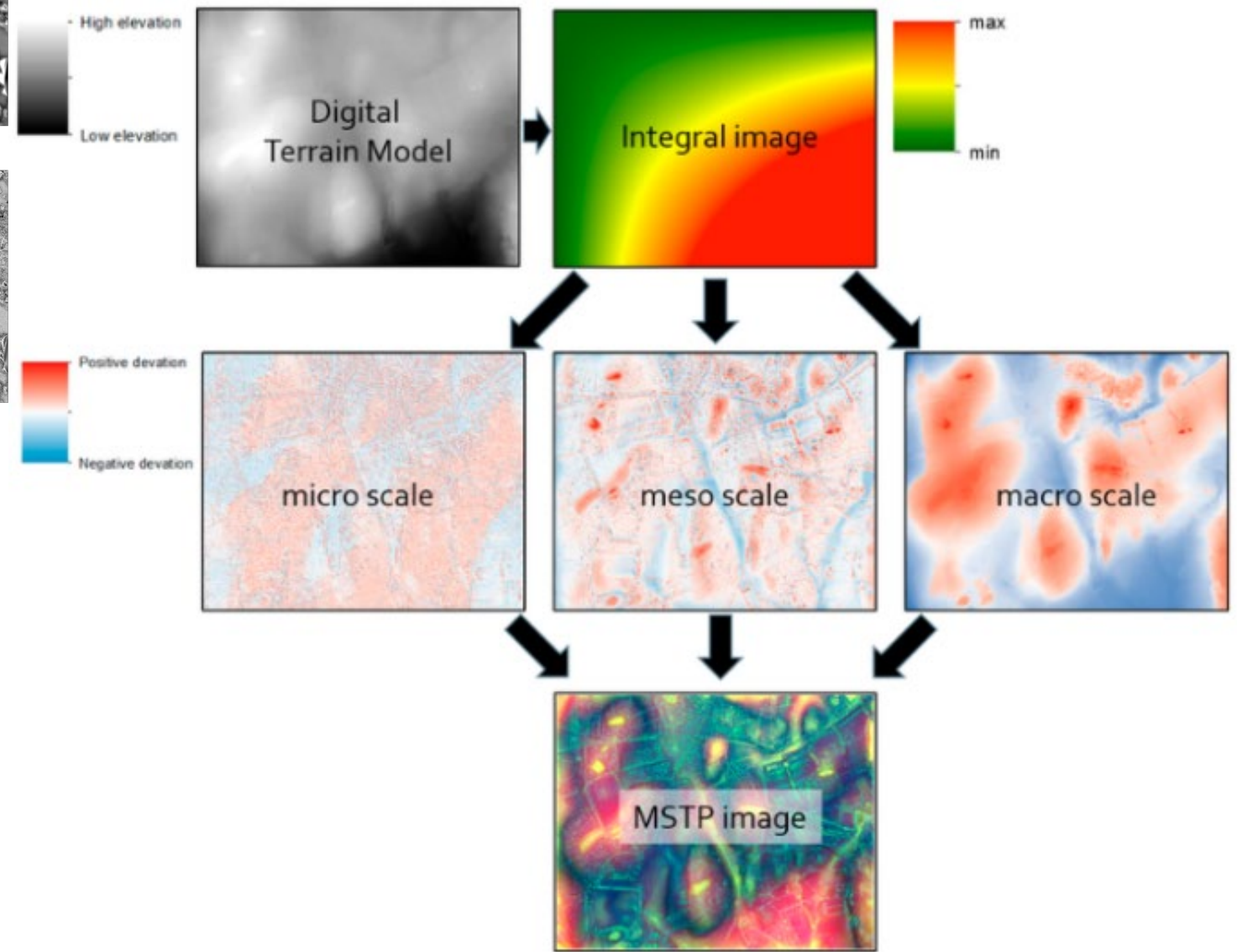
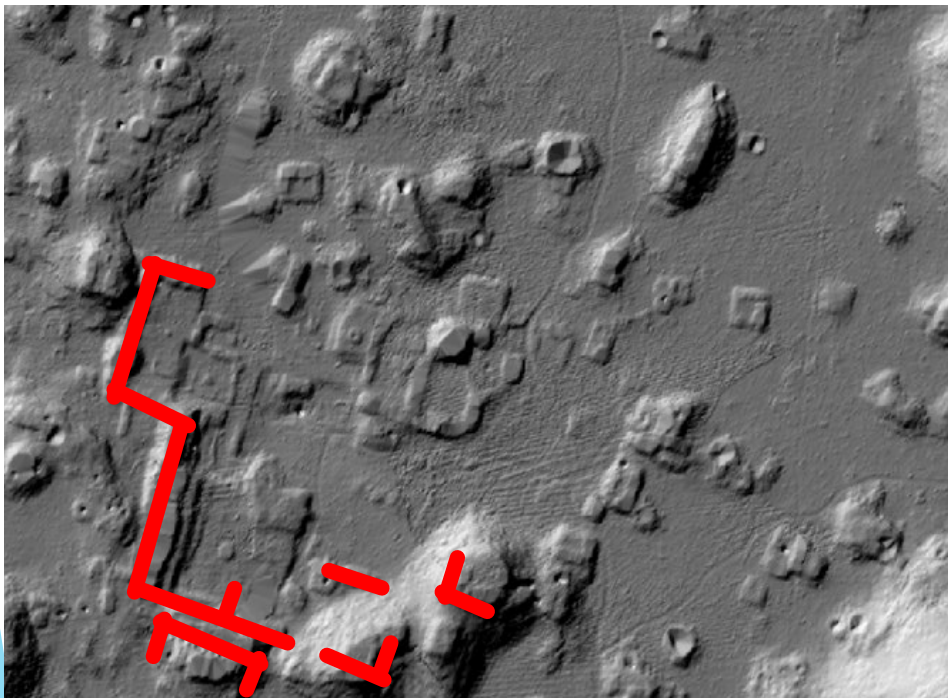
Simple local relief model



PCA of hillshading



Positive openness



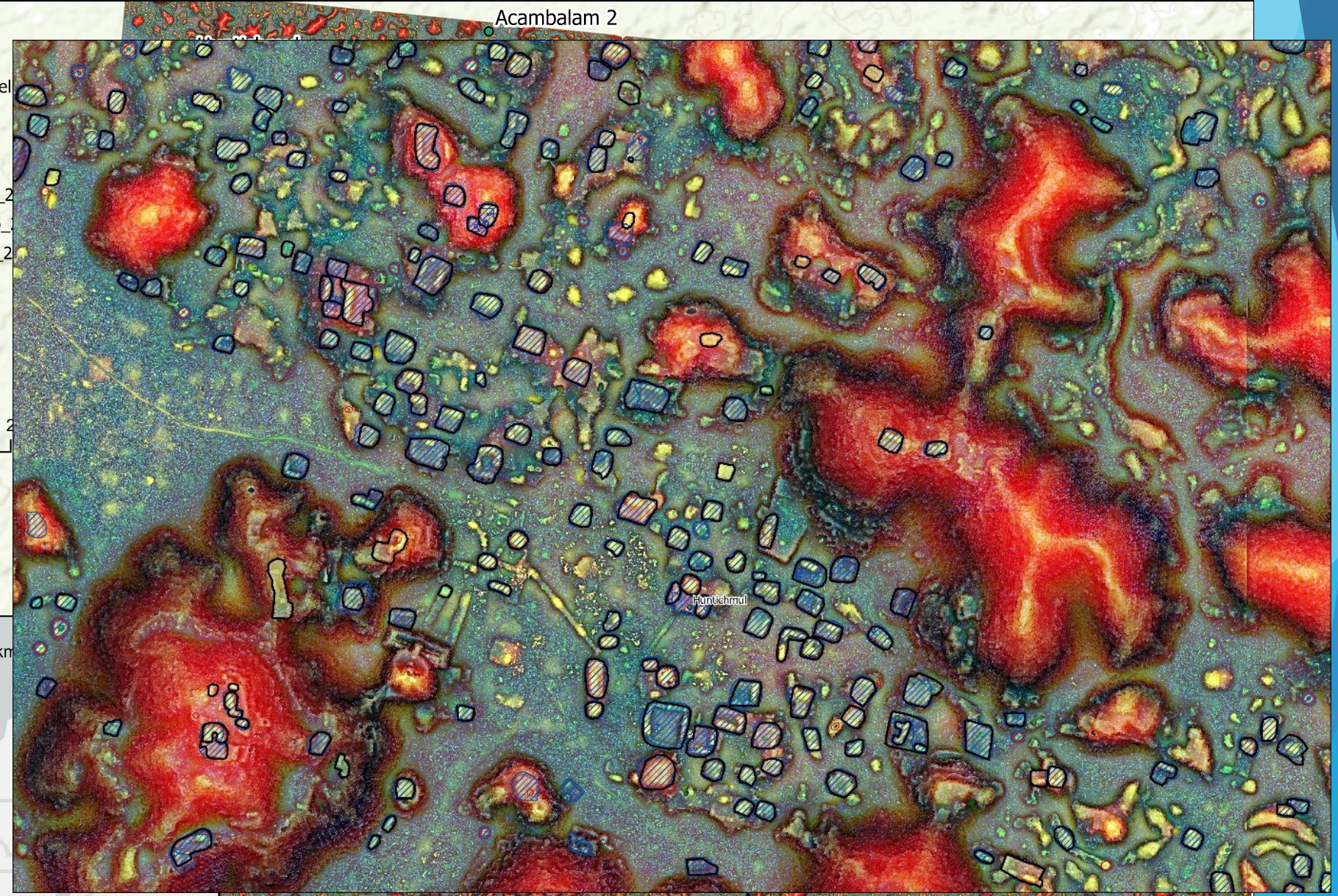
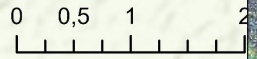
Adapted from "Combined detection and segmentation of archaeological structures from LiDAR data using a deep learning approach," door A. Guyot, M. Lennon, T. Lorho, & L. Hubert-Moy, (2021), Journal of Computer Applications in Archaeology, 4(1), p. 1. DOI: 10.5334/jcaa.64



# e3MSTP van het studiegebied verdeelt in 4 kwadranten

## Legende

- Maya-sites
  - Kwadrant verdel
- e3MSTP
- RGB
- Red: v3mstp\_2
  - Green: v3mstp\_1
  - Blue: v3mstp\_2

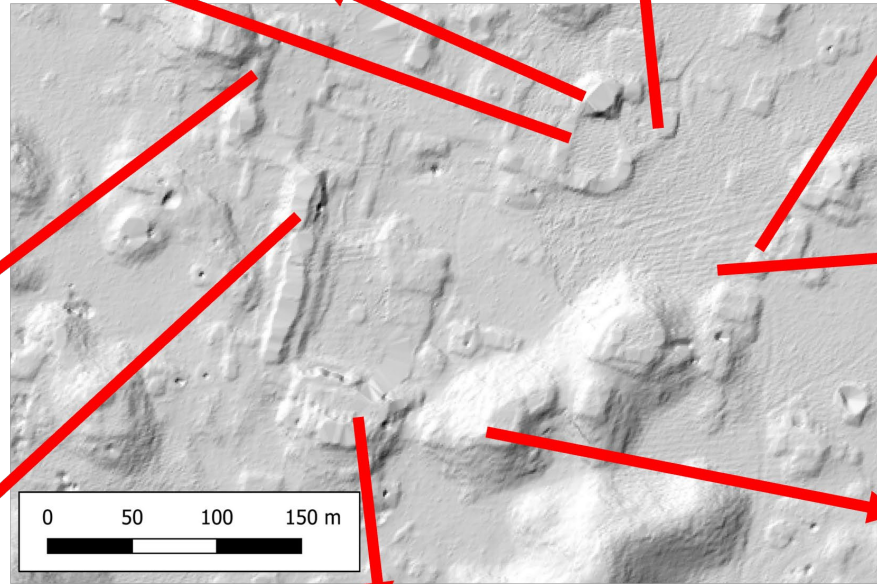


GUATEMALA

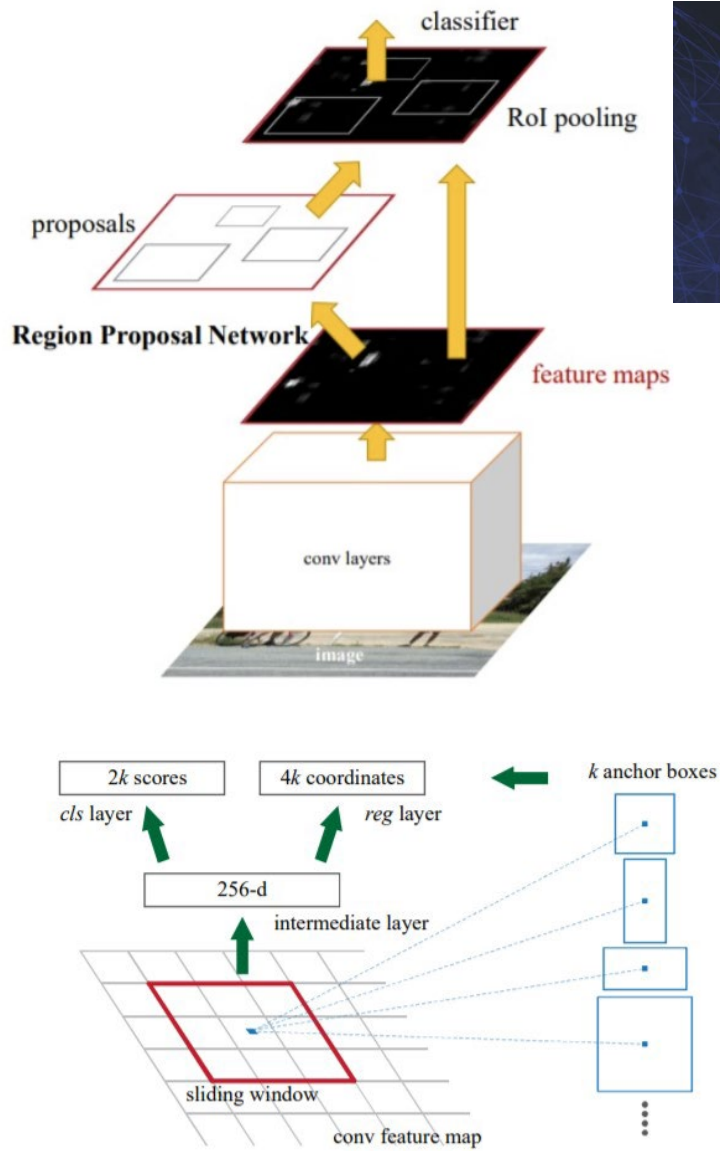
Guatemala, CONANP, Esri, TomTom, s  
Garmin, FAO, NOAA, USGS  
Tegucigalpa

Esri, HERE, Garmin, USGS, NGA, CPA, USDA





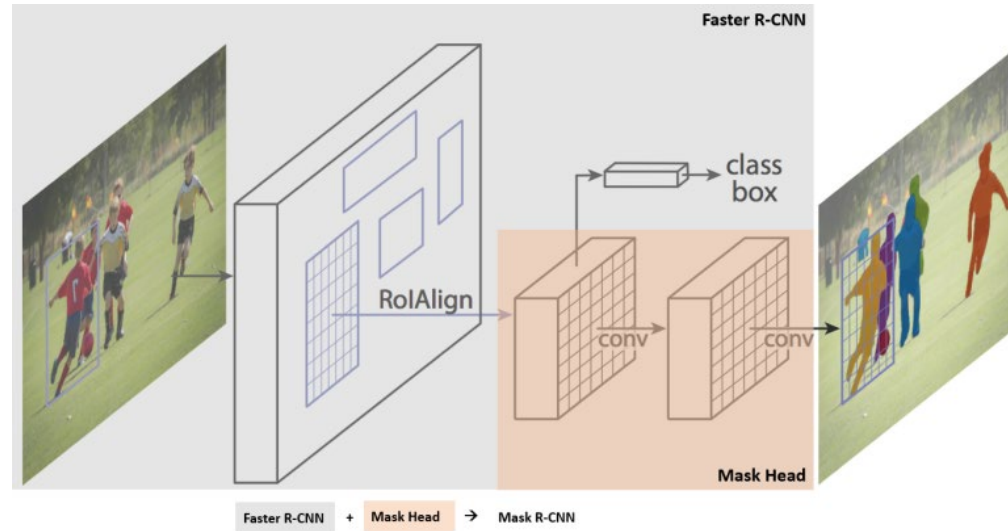




```
pc.fit(epochs=100, lr=myLR, early_stopping=True)
```

74.00% [74/100 1:48:40<38:10]

epoch	train_loss	valid_loss	average_precision	time
0	2.939283	2.338507	0.144679	01:51
1	2.139708	1.730215	0.322651	01:45
2	1.726137	1.465050	0.342089	01:32



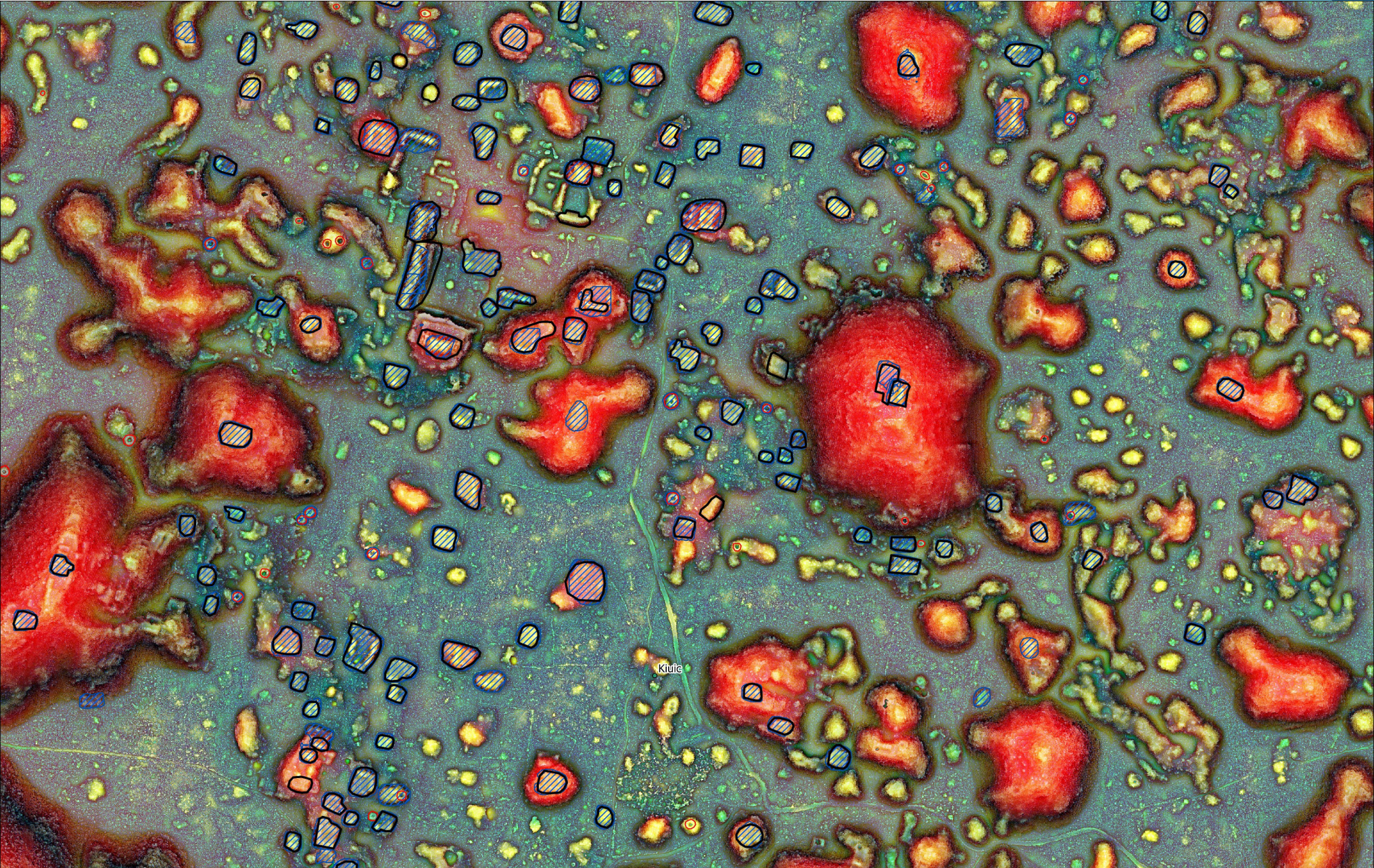
MASK-R CNN → Object instance segmentation

Adapted from "Faster R-CNN Object Detector | ArcGIS API for Python," (z.d.). Geraadpleegd via <https://developers.arcgis.com/python/guide/faster-rcnn-object-detector/>



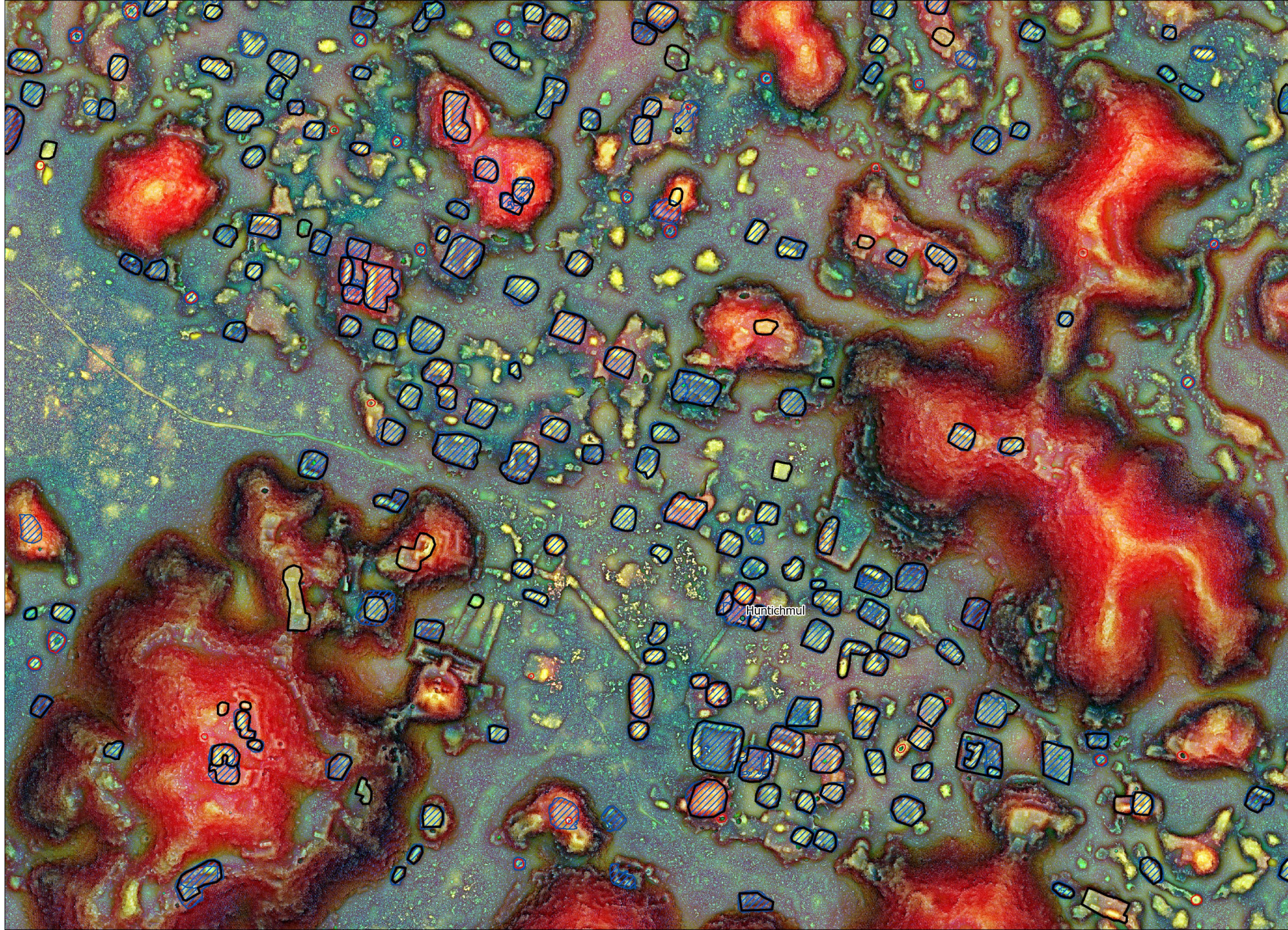
# Results

KIUIC





# HUNTICHMUL





- ▶ 6 → Buildings: good balance between precision and recall
- ▶ 1 → Ring-shaped structures: High precision but lower recall

IoU >= 0,3000

IoU >= 0,3000	Precision	Recall	F1 Score	AP	True Positive	False Positive	False Negative
All Classes	0,6696	0,8343	0,7430	0,6269	685,0000	338,0000	136,0000
6	0,6386	0,9213	0,7544	0,6731	585,0000	331,0000	50,0000
1	0,9346	0,5376	0,6826	0,5025	100,0000	7,0000	86,0000

## ▶ Post-processing

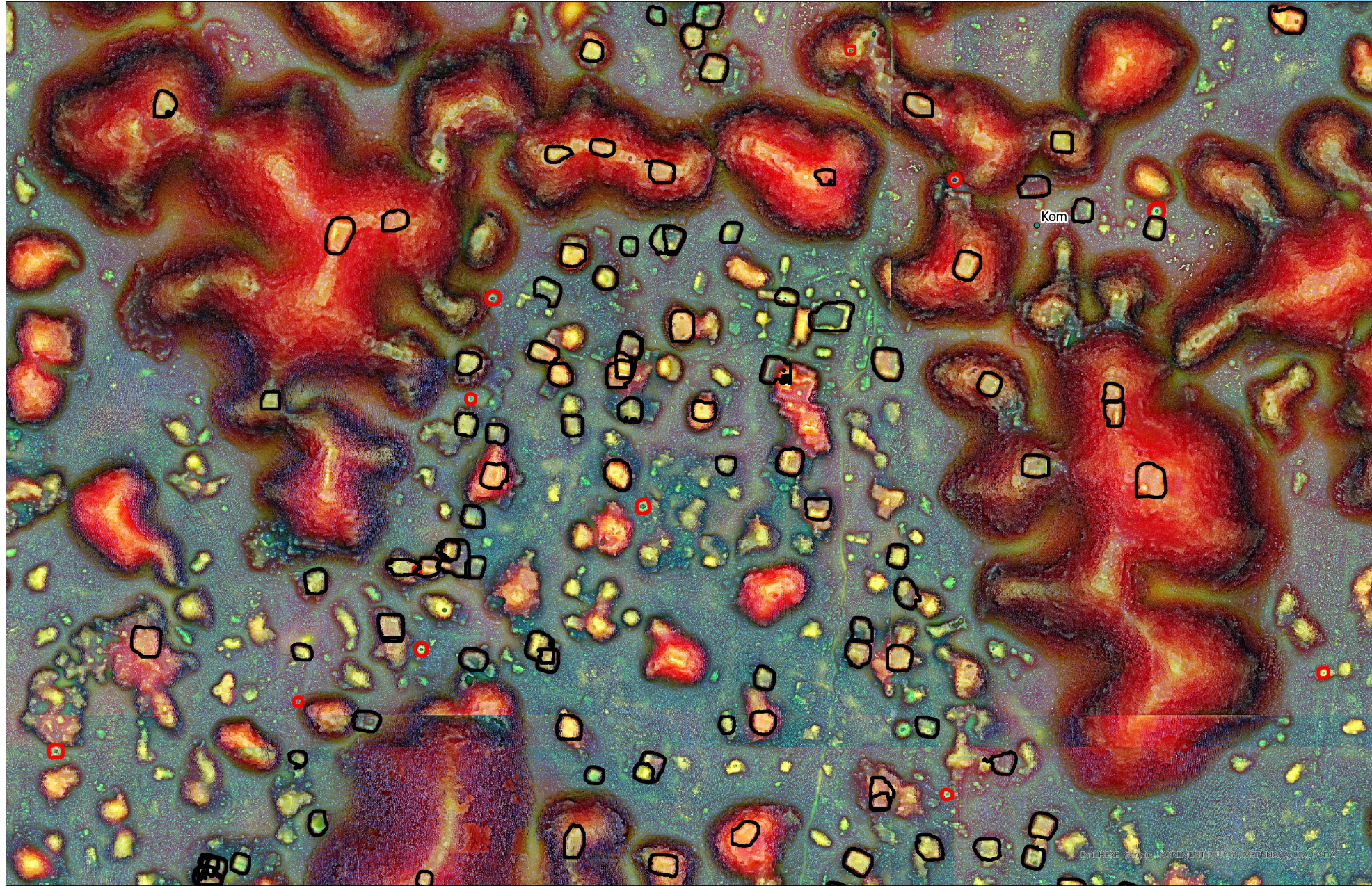
IoU >= 0,3000

IoU >= 0,3000	Precision	Recall	F1 Score	AP	True Positive	False Positive	False Negative
All Classes	0,7167	0,8343	0,7711	0,6559	683,0000	270,0000	136,0000
6	0,6891	0,9213	0,7885	0,7082	583,0000	263,0000	50,0000
1	0,9346	0,5376	0,6826	0,5025	100,0000	7,0000	86,0000



# Segmentation in "Unseen" Areas

KOM site







GROUND TRUTH  
MANUALLY ANNOTATED  
→ OBJECTIVE  
METHODS



EXPAND THE TRAINING  
DATASET FOR BETTER  
MODEL TRAINING



TEST DATASET



EXPLORE MORE DEEP  
LEARNING  
ARCHITECTURES AND  
BACKBONES



## **Buildings**

- Good balance between precision and recall
- Robust for building segmentation

## **Ring-shaped structures**

- High precision, low recall
- Model is cautious, few mistakes but many missed positives

## **Evaluation and Optimization**

- Current approach effective for identifying archaeological structures
- Further optimization needed to increase recall of ring-shaped structures

## **Key Insights**

- Accelerates the discovery and analysis of archaeological sites hidden under dense vegetation
- Enhanced techniques for cultural heritage preservation



