



- 1. DEEP LEARNING
- 2. CUSTOM GEO-INFORMATION
- 3. USE CASES



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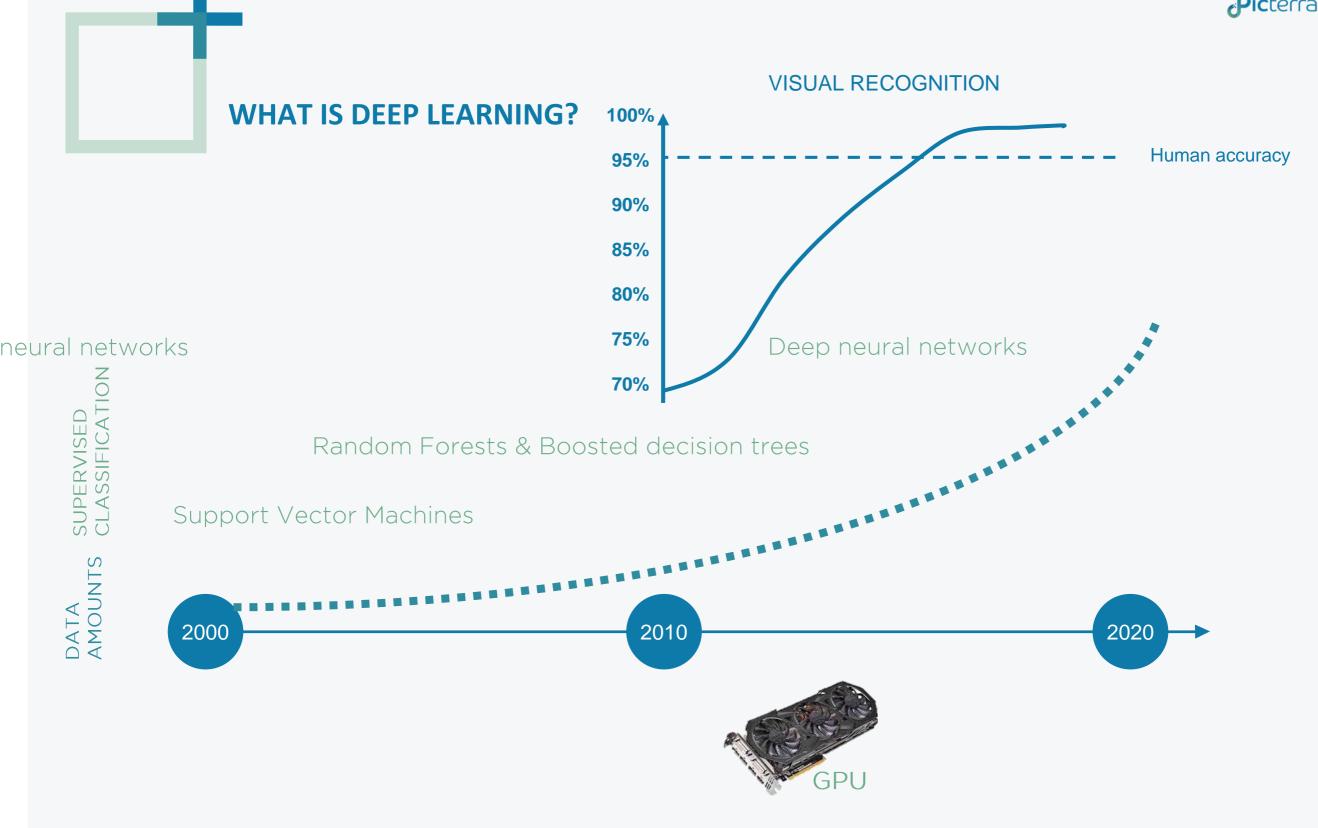


### MACHINE LEARNING

### SUPERVISED CLASSIFICATION

TRAINING DATASETS

COMPUTATIONAL POWER







#### WHAT IS DEEP LEARNING?



Hand-crafted features

- Colour histograms, gray co-occurrences, morpholgy
- Histogram of Orientated Gradients (HOG)
- corners and edges detectors (SIFT, FREAK, BRIEF, etc.)
- Bag of Visual Words (BoVW)

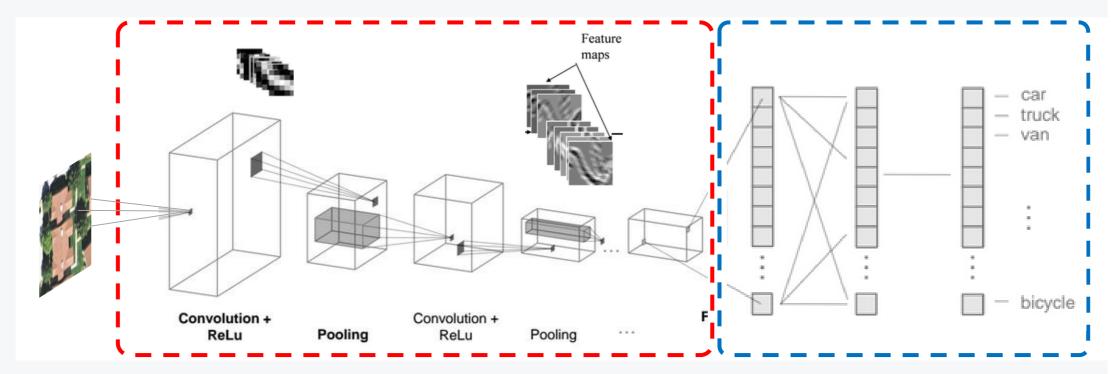
Random Forests

Boosted decision trees

Support Vector Machines

Feature extraction

Classification



Deep learning == end-to-end learning



# WHAT IS DEEP LEARNING?

# Semantic segmentation

### Object detection

# Object mask detection



Every pixel get a thematic label



Only coordinates of bounding boxes



Every pixel in bounding boxes get a thematic label

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**Product Manager** 

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JULIEN REBETEZ Lead Software & Machine Learning

LÉO ROCHER Full stack engineer





























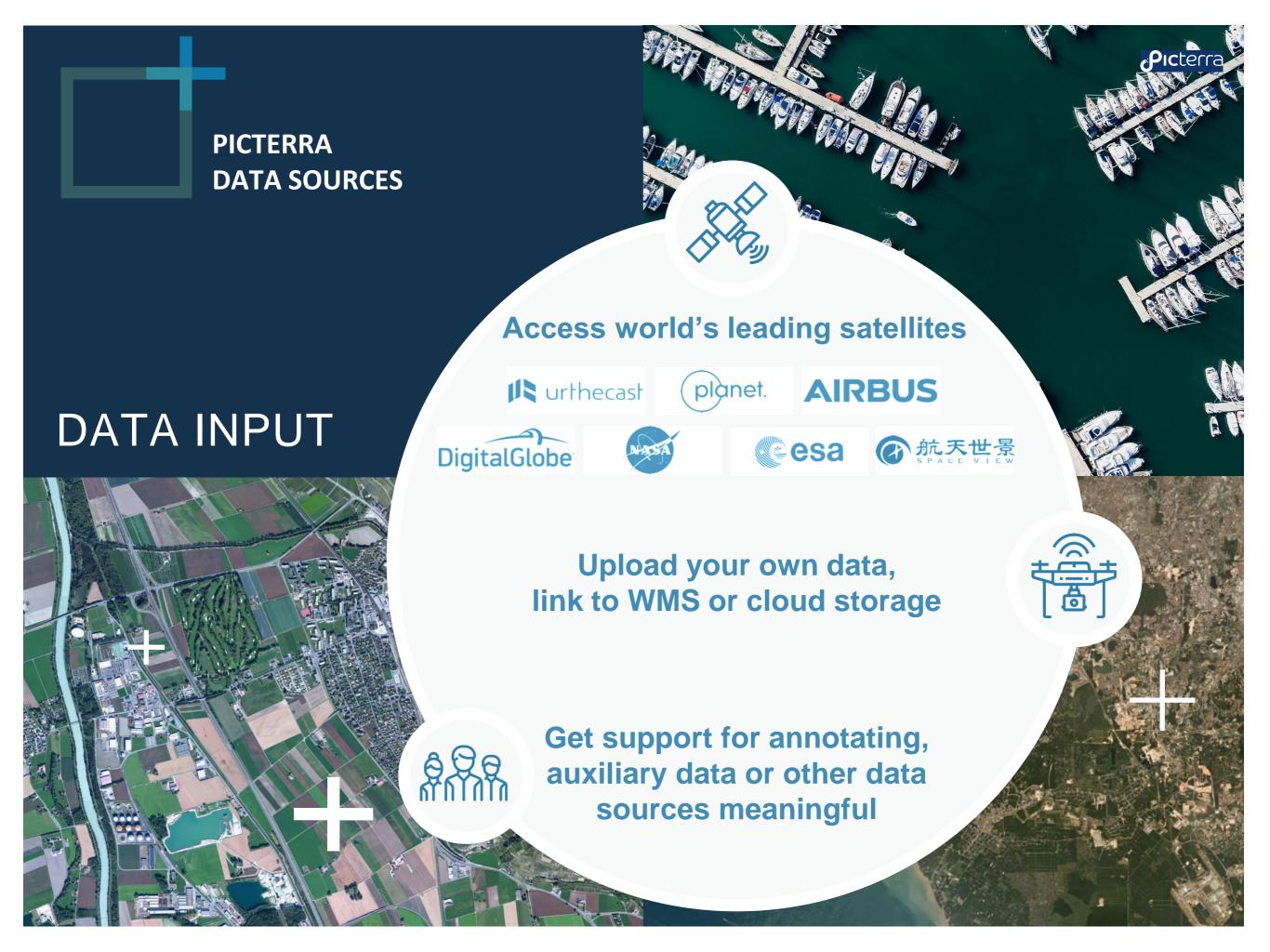














### PICTERRA SOLUTION

SELF-SERVE
TOOL OFFERING
USERS ACCESS
AND CONTROL

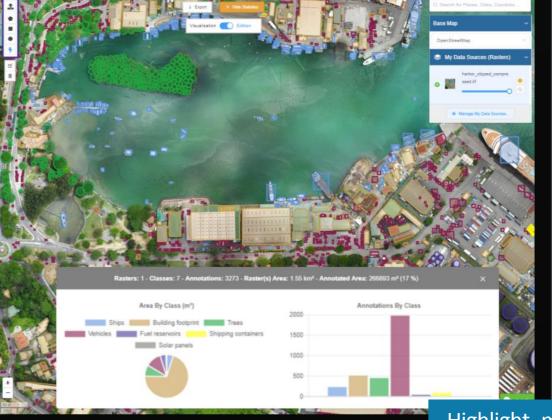
Classes

Classes

Chapter Classes for your objects of pricests. Scient a Class and a Dreating Tool and start to Annotate your maps Tool and Scientific Tools Tools

Draw and categorize annotations into different classes

Annotate faster by using Al powered tools, which automatically detects similar objects





Highlight, measure, count objects and share your projects with others









# FROM DEEP LEARNING TO CUSTOM GEO-INFORMATION

What does this people have in common?

















Detect and outline "objects" in imagery



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#### PRECISION AGRICULTURE





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#### PRECISION AGRICULTURE

Detections

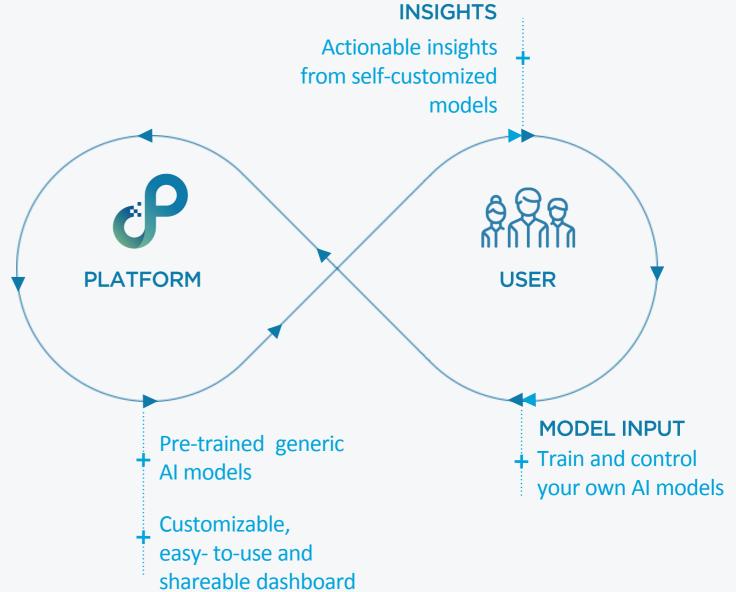








- + Users can leverage their proprietary data or tap into Picterra's network of trusted data providers
- + Satellite imagery available from 30m to 30cm resolution



GENERATE CUSTOMIZED

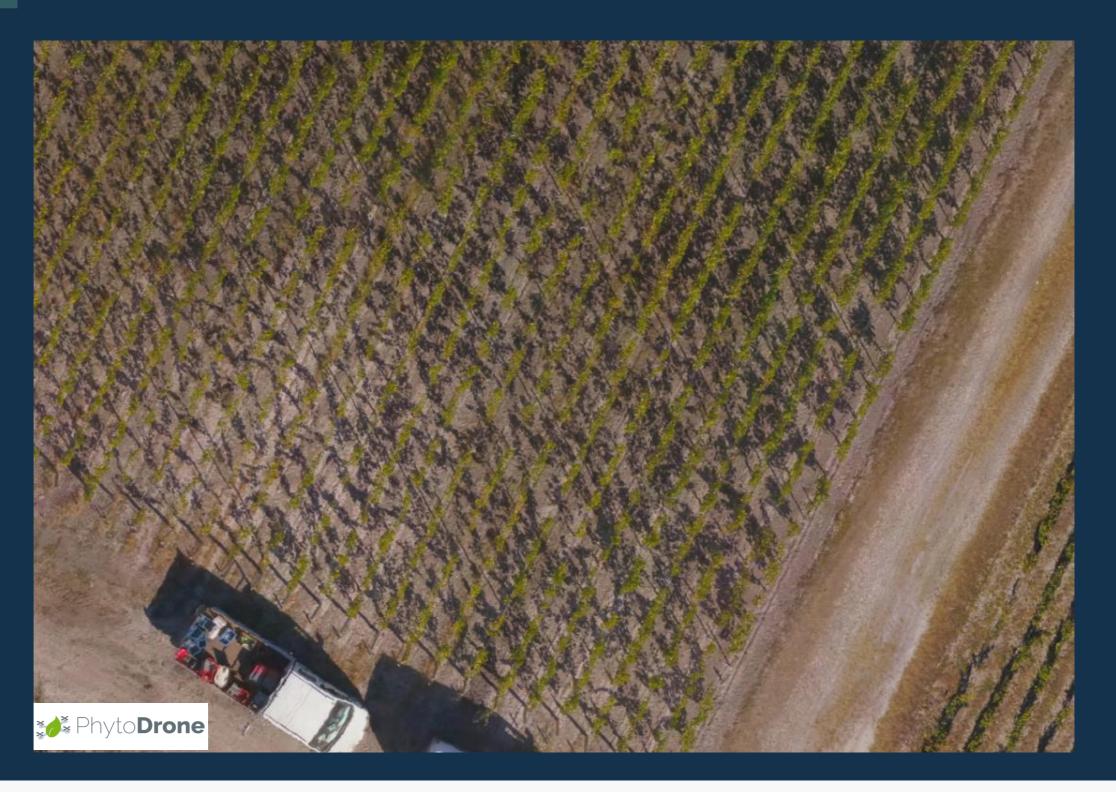
GEO-SPATIAL

INFORMATION





### PRECISION AGRICULTURE







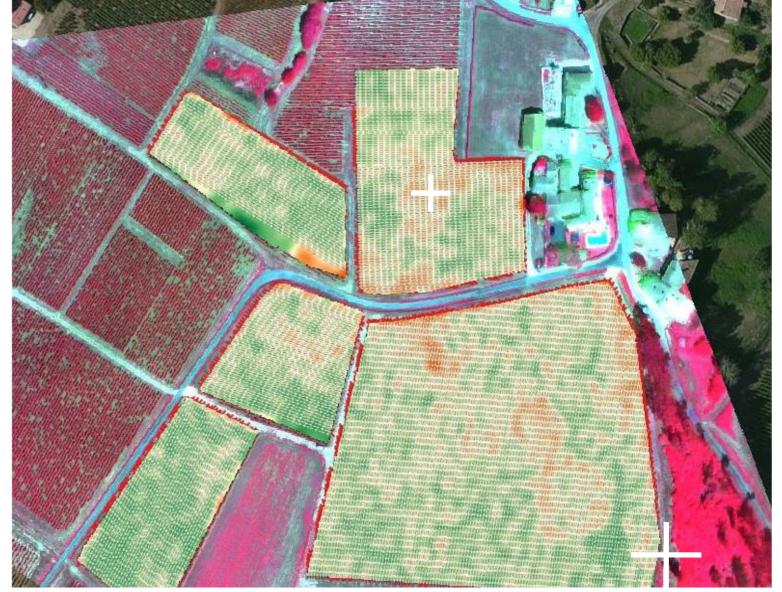
### PRECISION AGRICULTURE





#### PRECISION AGRICULTURE





#### Parcel statistics

- ▲ Number of rows: 40
- ▲ Orientation of rows: East (107°)
- ▲ Number of vines: 1924
- ▲ Number of missing vines: 493
- ▲ Vine density: 5100 vines/ ba







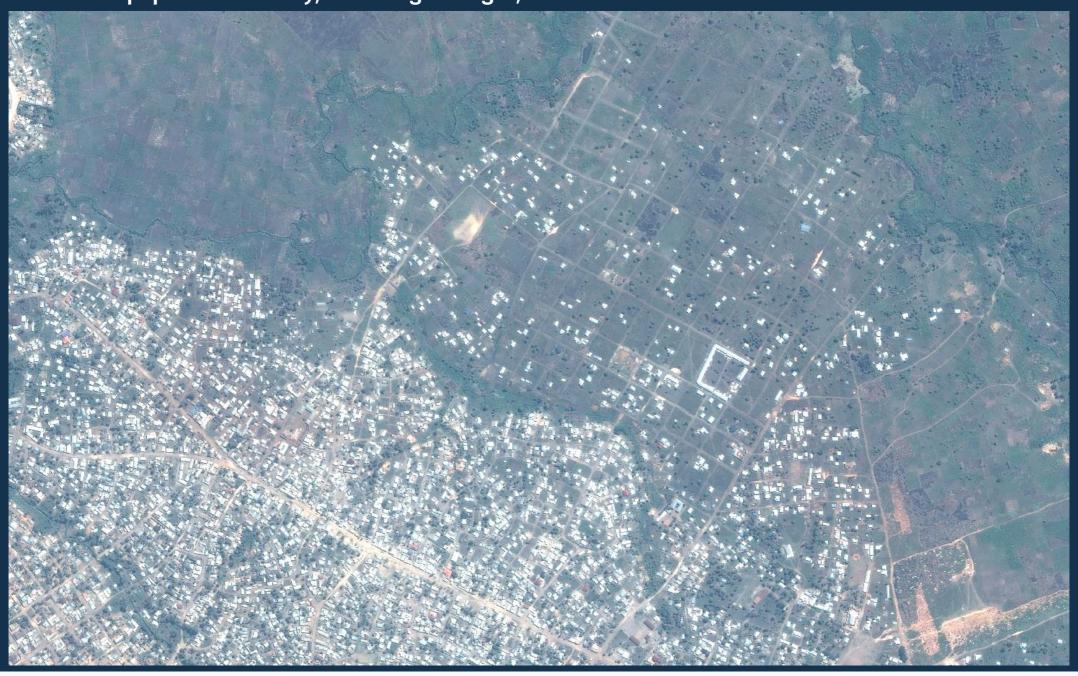
#### URBAN & BUILDING FOOTPRINT

Detecting vehicles in urban environment: parking usages, mobility planning



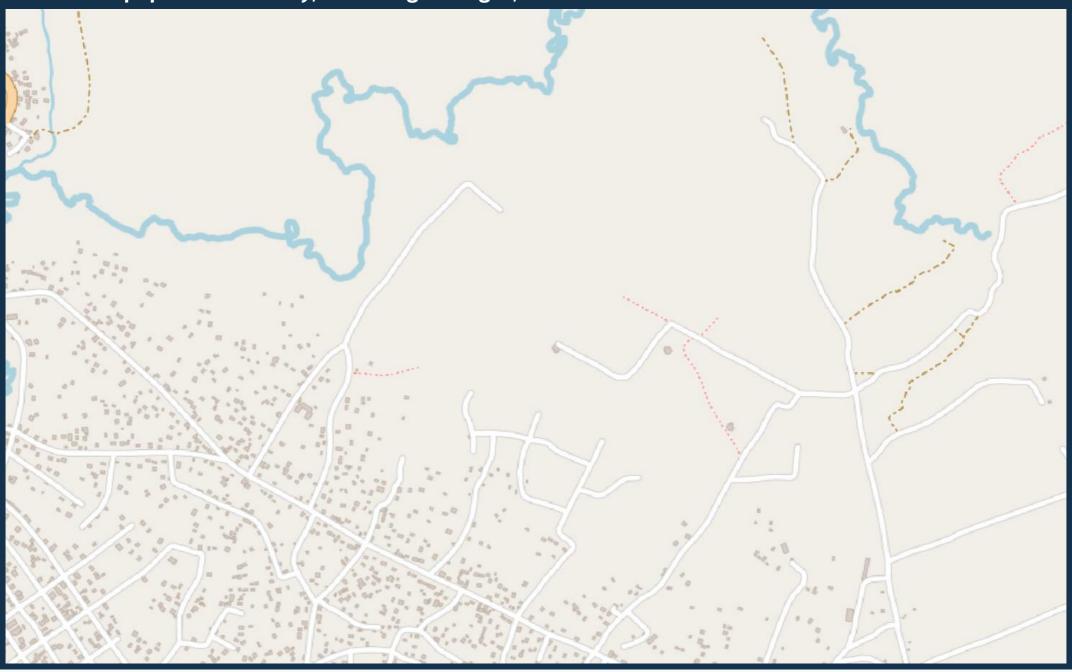


### URBAN & BUILDING FOOTPRINT





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#### URBAN & BUILDING FOOTPRINT





#### URBAN & BUILDING FOOTPRINT

#### Solar panels detection over building roofs





#### LIVESTOCK MANAGEMENT

#### Fast livestock management for counting populations of animals





#### LIVESTOCK MANAGEMENT

#### Fast livestock management for counting populations of animals





#### LIVESTOCK MANAGEMENT

#### Fast livestock management for counting populations of animals

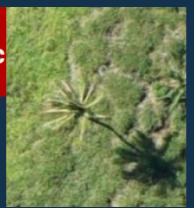








Open AI Challenge for the South Pacific





Flying Labs



# ETHzürich

Winning team: Karla Saldaña and Guo Zifeng

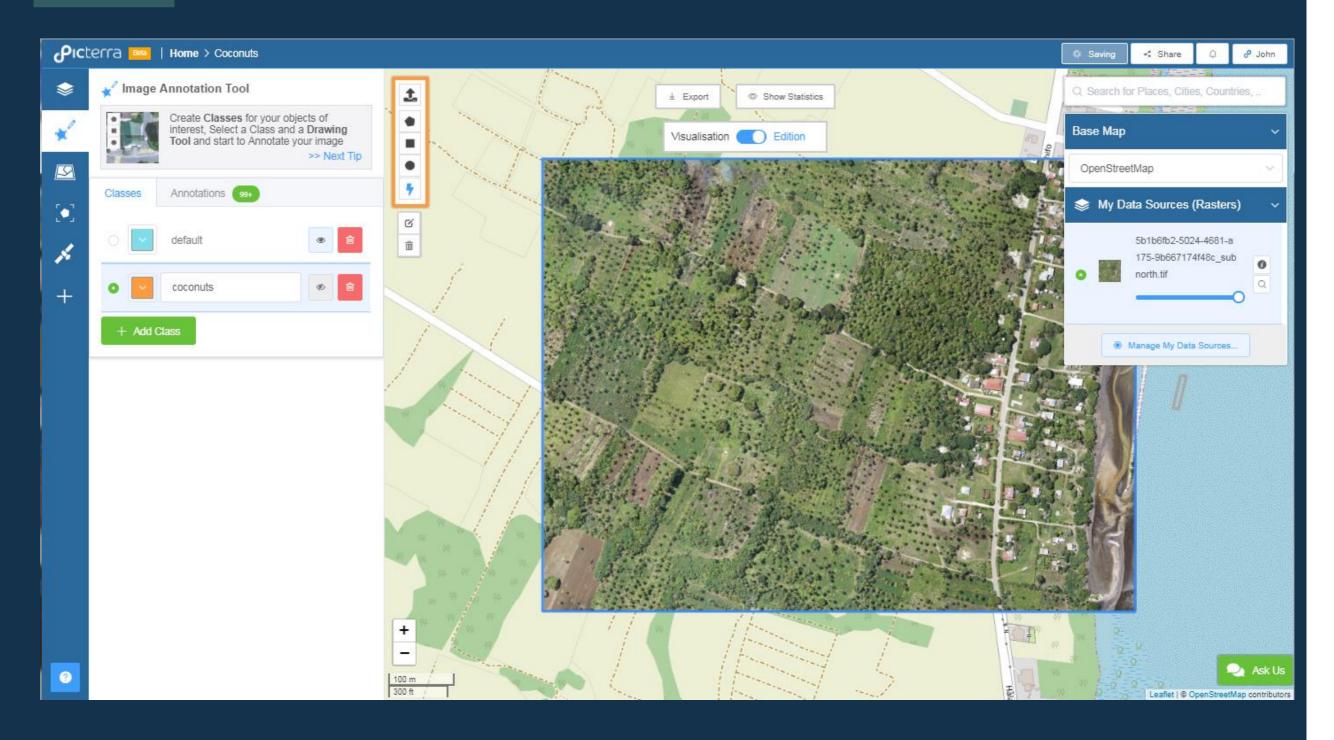




Integrated

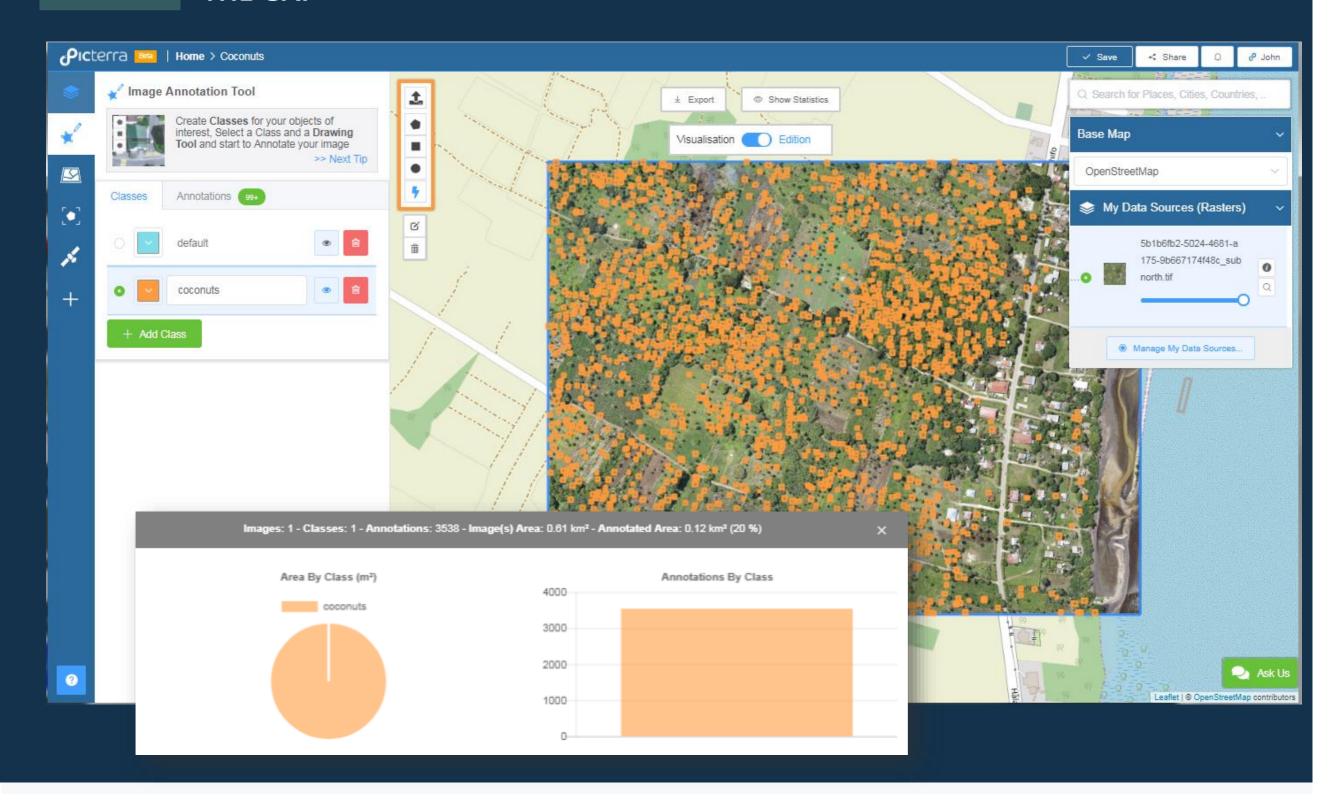


# BRIDGING THE GAP





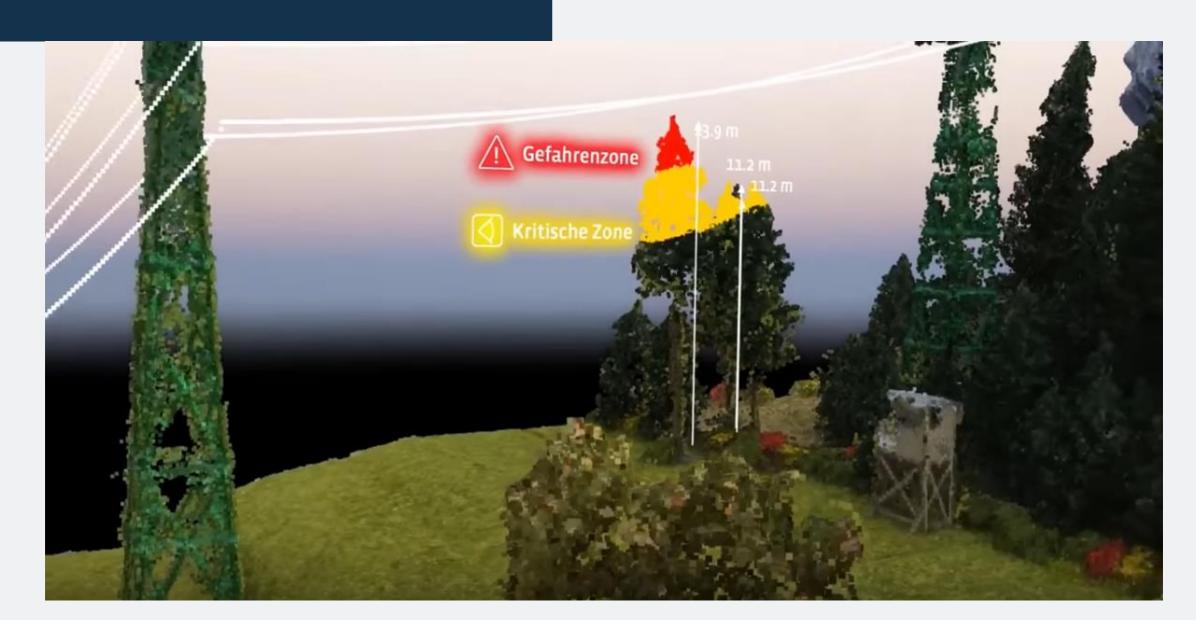
# BRIDGING THE GAP





#### OUTLOOK

WHAT ABOUT DEEP LEARNING
ON 3D POINTCLOUDS?







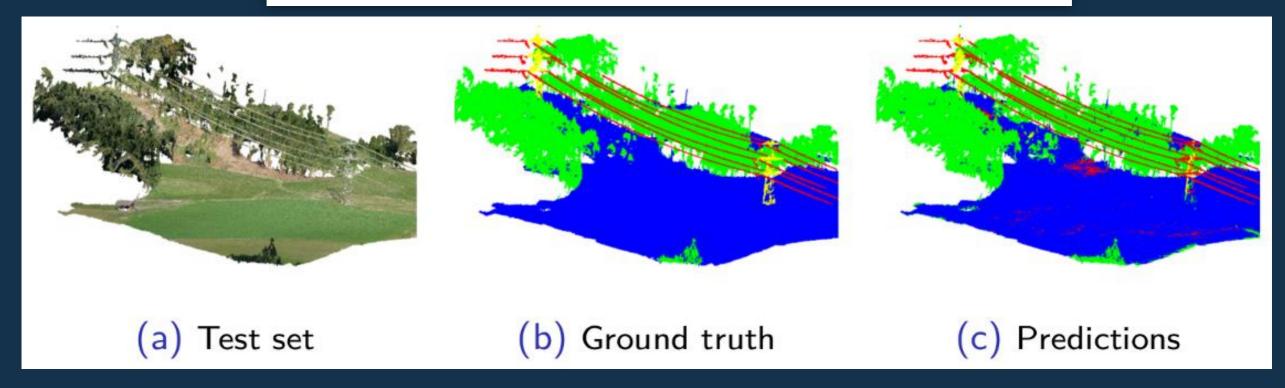






#### Deep learning on 3D pointclouds – Learning on data graphs

Performances	Overall accuracy (in %)	Mean accuracy (in %)
Random Forest	71.32	43.57
XGBoost	75.34	52.86
Our model	95.15	84.07
Majority class	54.66	25.00







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#### LAND COVER & USE CLASSIFICATION

#### Land cover and land use statistics can benefit from several data streams

