

# Alteia

Agriculture Machine Learning

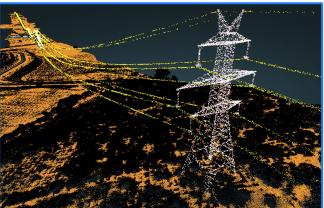
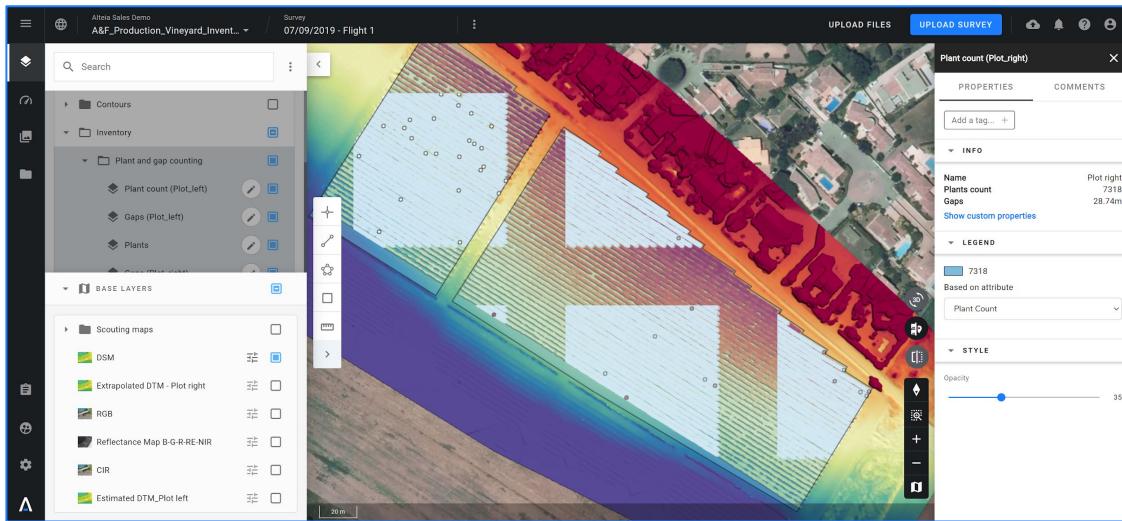
## Who am I



- Head of Data science team @ [Alteia](#) 
- Passionate about Deep learning and math in general 
- You can contact me here: [Linkedin](#) 

## MANAGE PHYSICAL ASSETS - 2D•3D•4D

- Run Analytics from satellite
- Extract Insights from drone mapping
- Model Reality from air & ground scanners
- Inspect & Detect from mobile devices

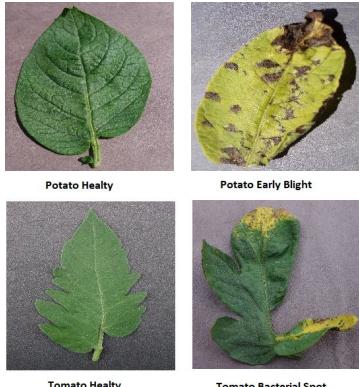


Big Data et Agriculture

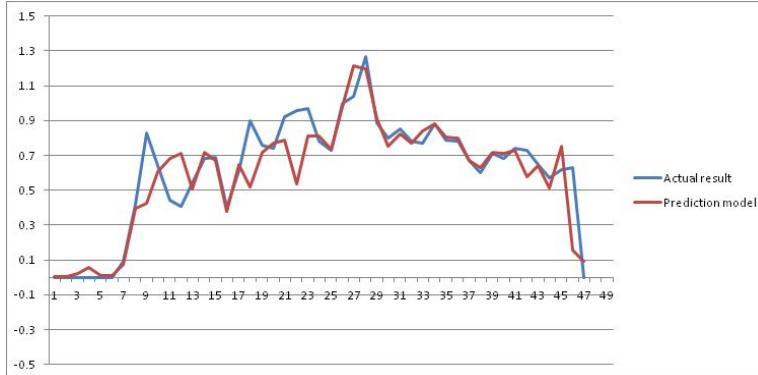
19/10/2021

Alteia - 2021

# Agriculture and Data Science



- Yield prediction
- Disease detection
- Weed detection
- Crop recognition
- Crop quality



- Generalities of Machine Learning
- Data acquisition methods and data sources
- Case study
- Conclusions and Q&A

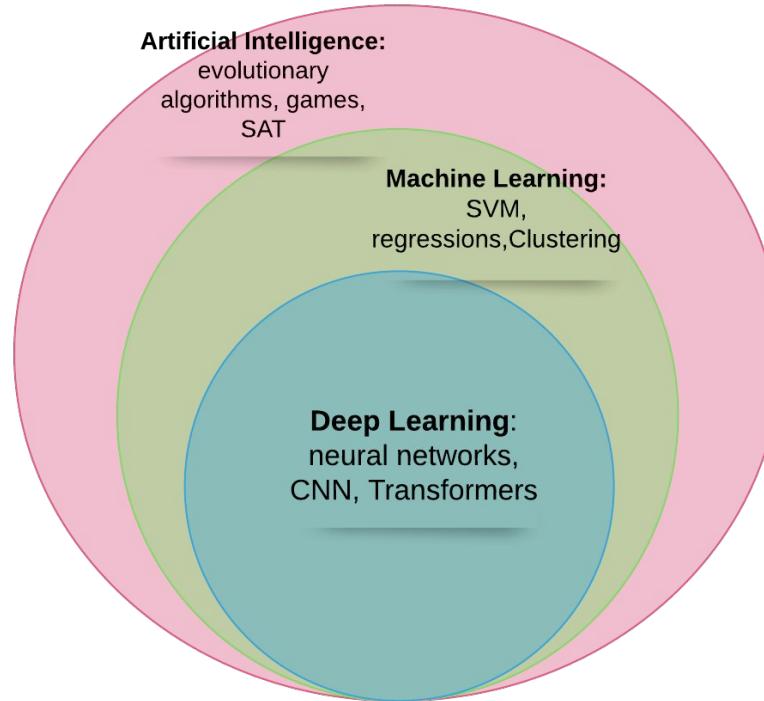
# Machine Learning

## Artificial Intelligence de-buzz the buzz word

AI is just a set of techniques to solve some technical problems.

It does not mean that is “intelligent” (what does it even mean ?)

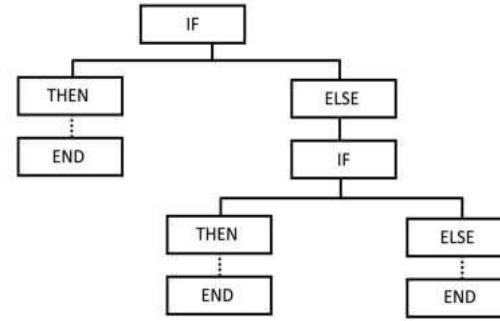
No robot army will rise in the near future. It is difficult enough to solve simple problems without any need of consciousness.



## Machine Learning Programming model

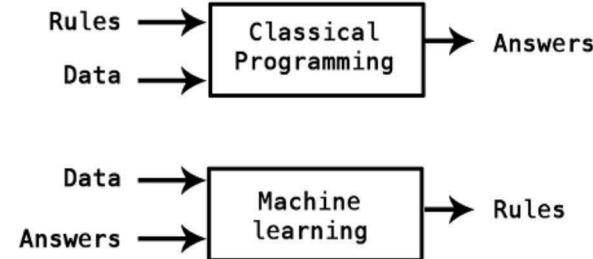
**Rule-based:** rules are explicitly given to the system

- Aims for decision-making
- Example: AIs in video games
- Con: Difficult to build such AI for complex problems

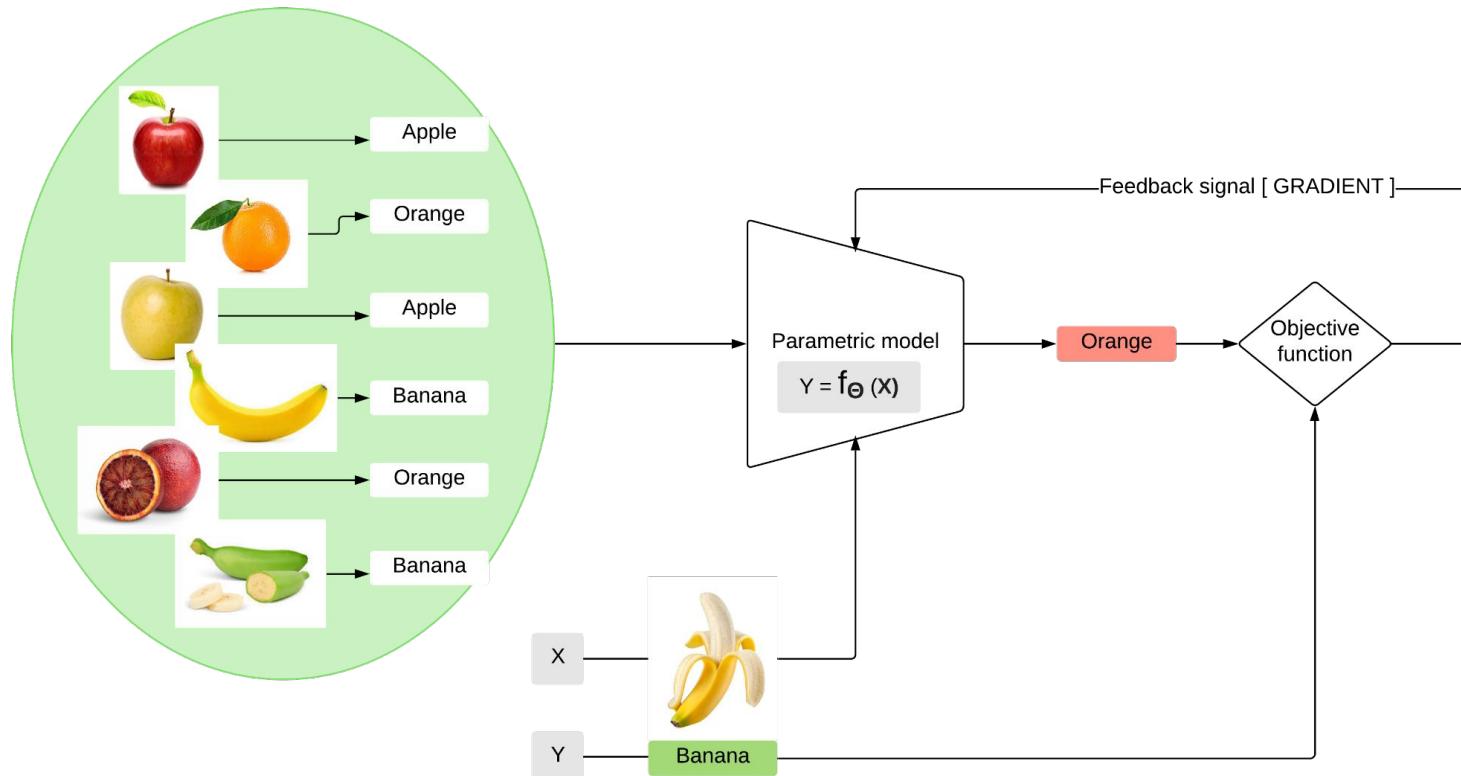


**Data-based:** problem inversion, the system learns the rules

- Finds patterns in the data
- Needs A LOT of data (Big Data)
- Can be hard to explain the outputs (black-box model)
- Machine learning is data-driven
- May have unpredictable behaviors



## Key concepts in machine learning model training



## Different paradigms

- Supervised Learning:

- A data point is defined by its features and its label
- Training on *labelled* data, predictions on new unlabeled data

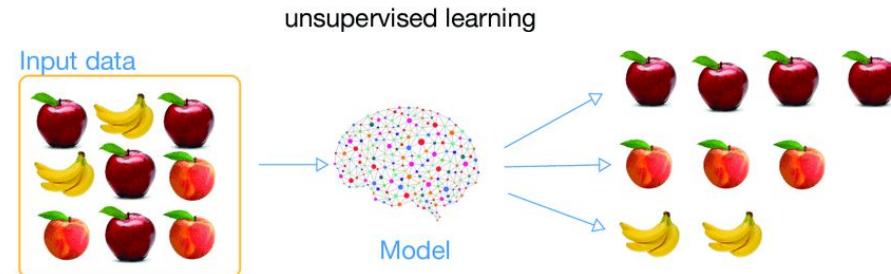
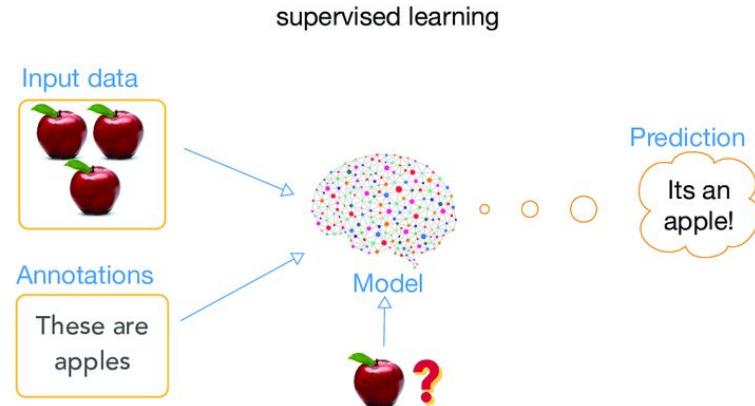
- Unsupervised Learning:

- A data point is defined by its features only
- To separate *unlabeled* data into different categories

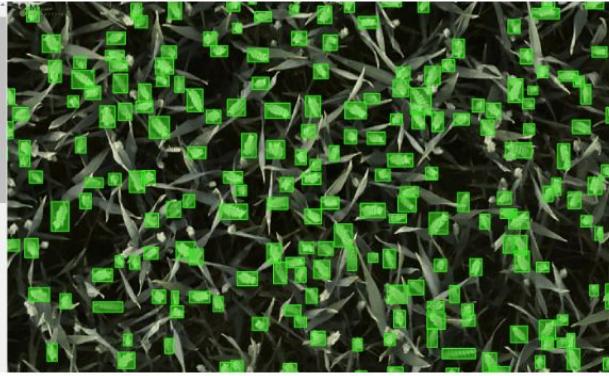
These are two different ways to solve the same problem.

The difference lays in the accuracy that we can get and the budget of data that we have.

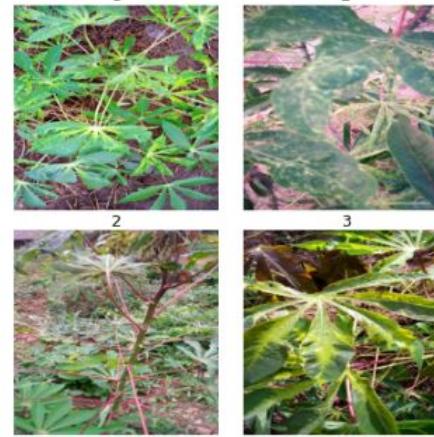
Using unsupervised learning is less data efficient (usually).



## Agriculture tasks and problems

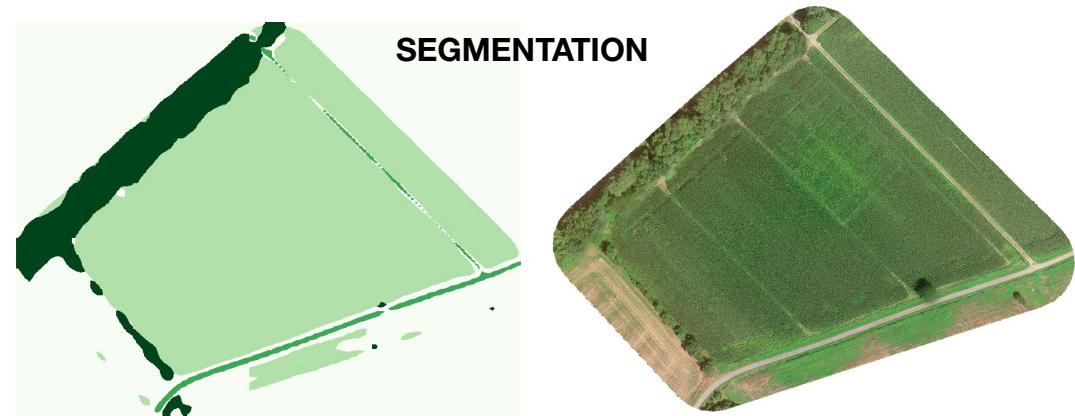


**DETECTION**

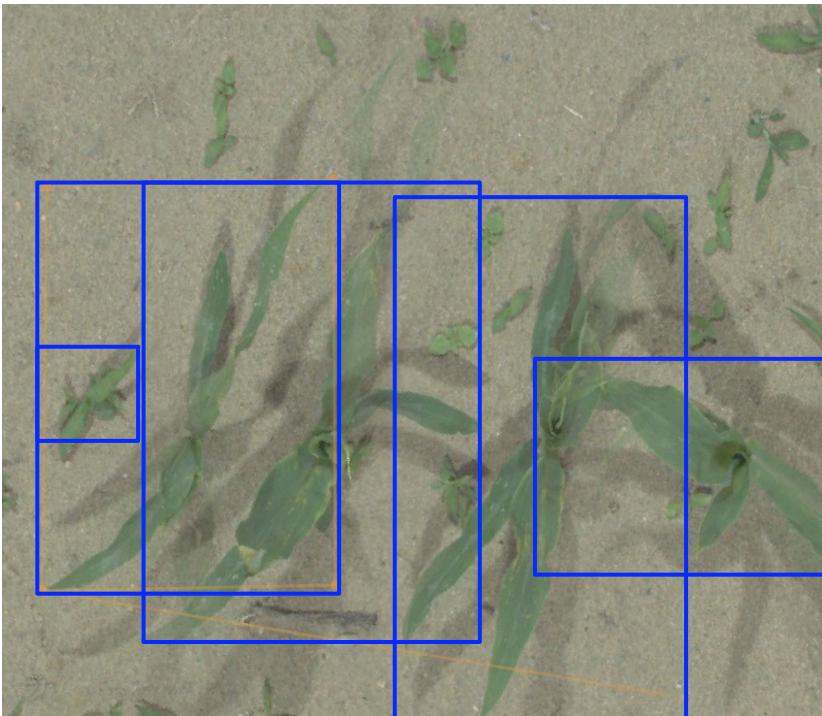


## CLASSIFICATION

<https://www.kaggle.com/c/cassava-leaf-disease-classification/data>

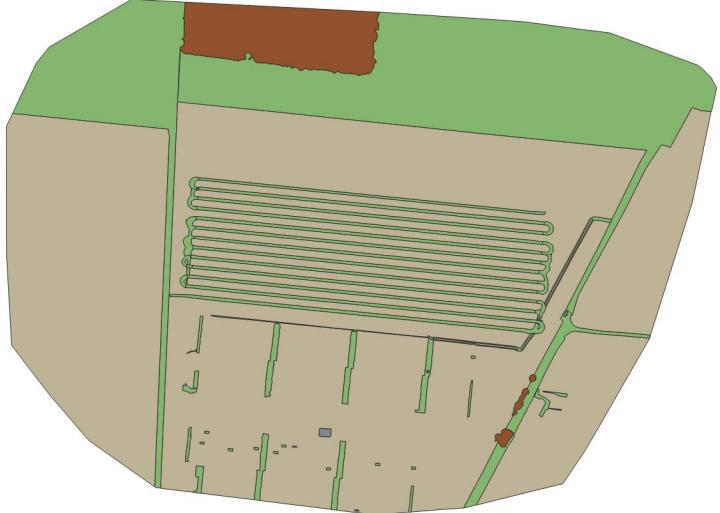
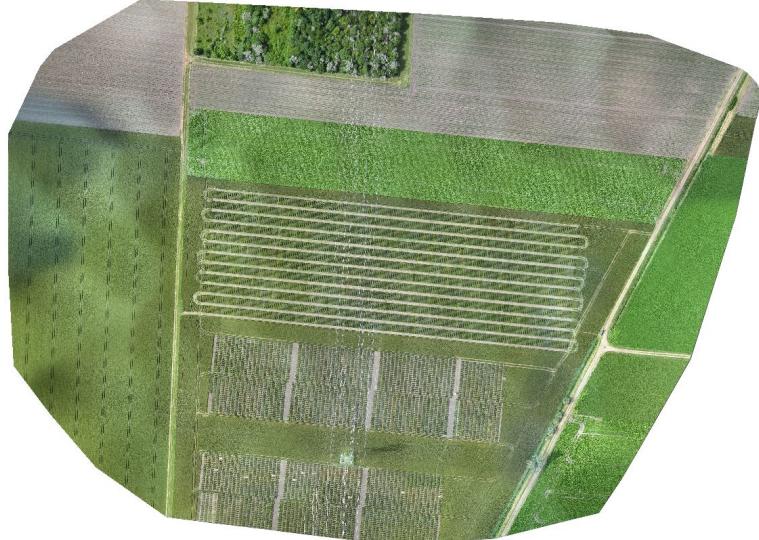


## Importance of Labelling



## Constraints of Labelling

- **Time:** do you need to label a dataset fast
- **Quality:** can you afford noise and errors in the dataset
- **Cost:** how much budget do you have
- **Location:** can you subcontract the label process
- **Type:** bounding box are easier than polygons



# Data Acquisition

## Remote sensing, AKA drones, robots and satellites

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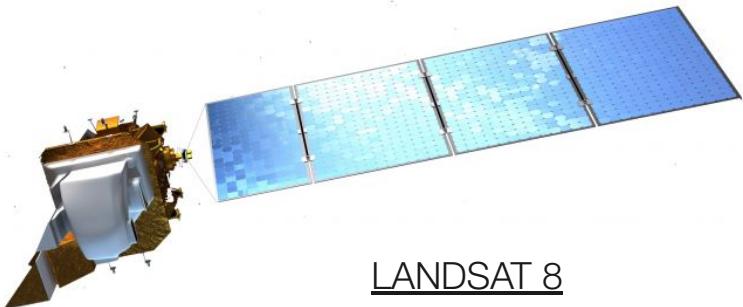
DELAIR



DJI



NAIO



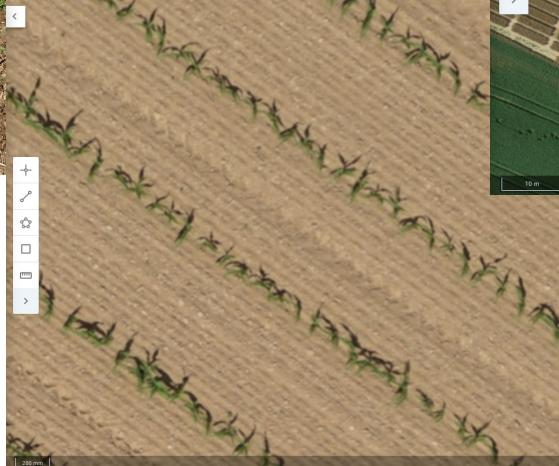
LANDSAT 8

## Spatial Resolution Importance



~ mm

~ 1 cm



~ 5 cm

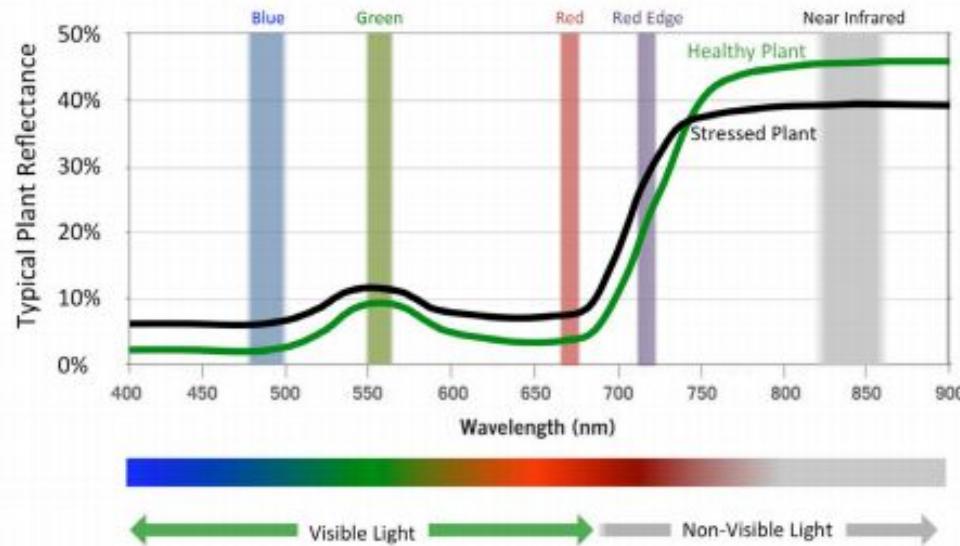
~ m



## Cameras and sensors

### Spectral bands

High-grade optical filters deliver precise information specially targeted to agricultural applications.



Micasense

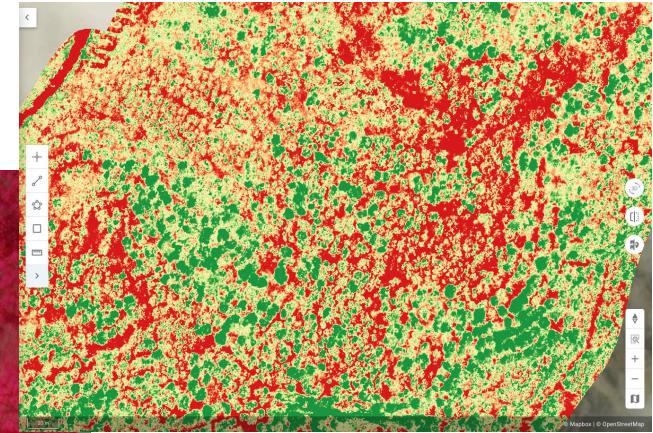
## Different kinds of data for different problems



RGB

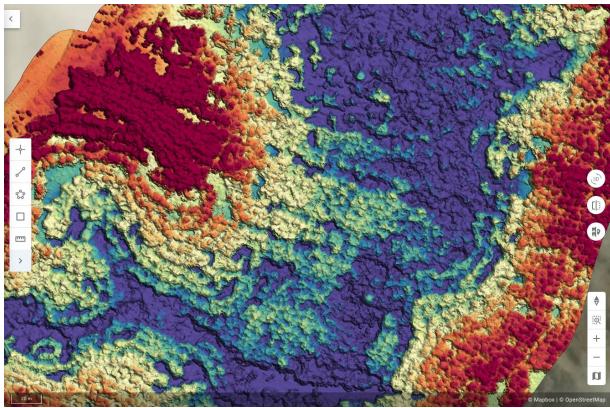


CIR

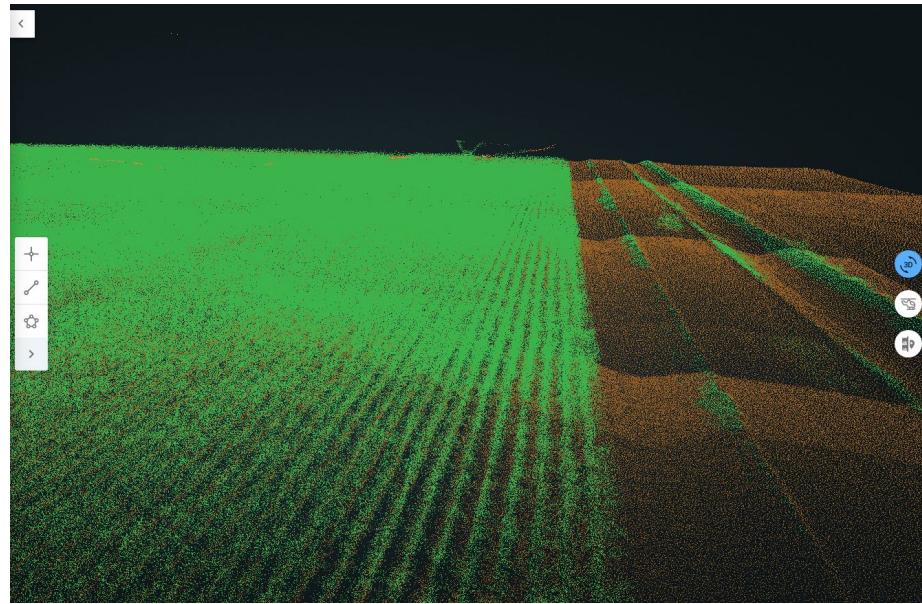


NDVI

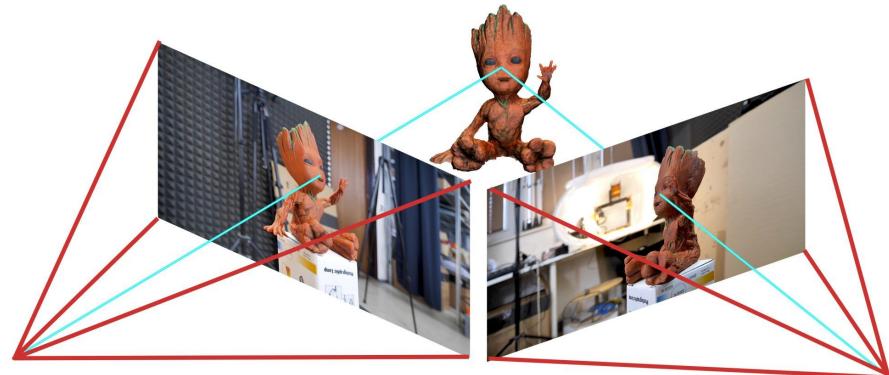
## 3D Height Data



DSM



LIDAR



## How much data do I need ?

### BIG DATA is just a buzz-word:

You should ask yourself how much data do you need for a certain application and how much data you produce in a certain time frame; then select the solution of storage and retrieval accordingly

#### CLOUD



##### Pro:

- You do not manage bare metal
- You can pay for what you use
- Adapt to peaks and lows
- Always last technology

##### Cons:

- Can be difficult to setup
- Do not work offline
- Data retrieval speed can be slow

#### LOCAL



##### Pro:

- Works offline
- Data retrieval is faster
- Security (privacy politics)

##### Cons:

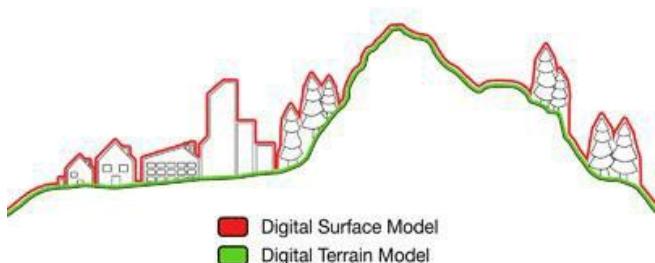
- You need to manage machines (install, update etc...)

# CASE STUDY 1

## Plant Height Estimation

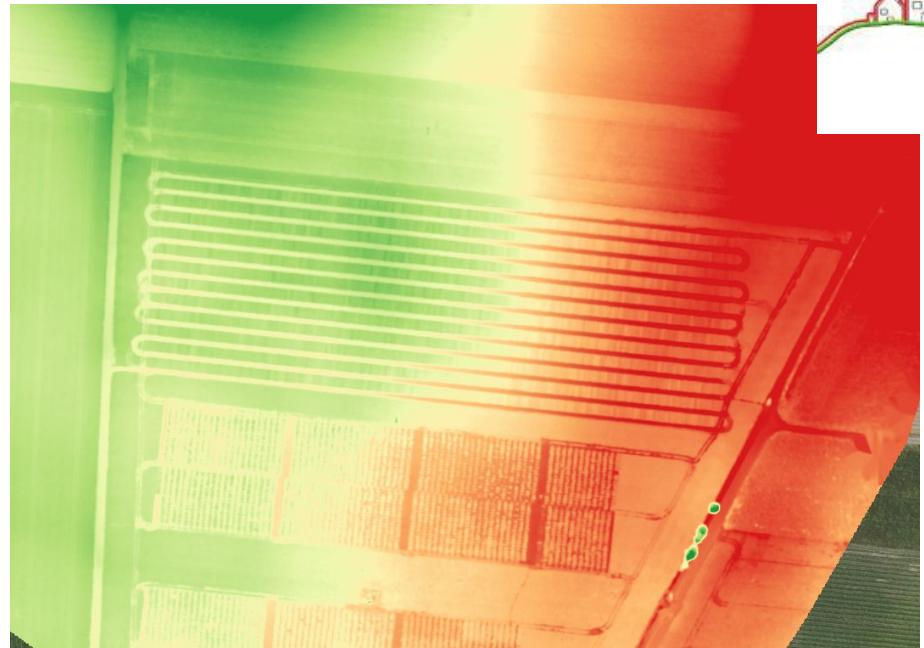
## Objectives

- Estimate the height of the vegetation

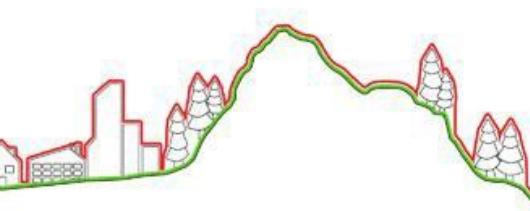


## Dsm - Dtm

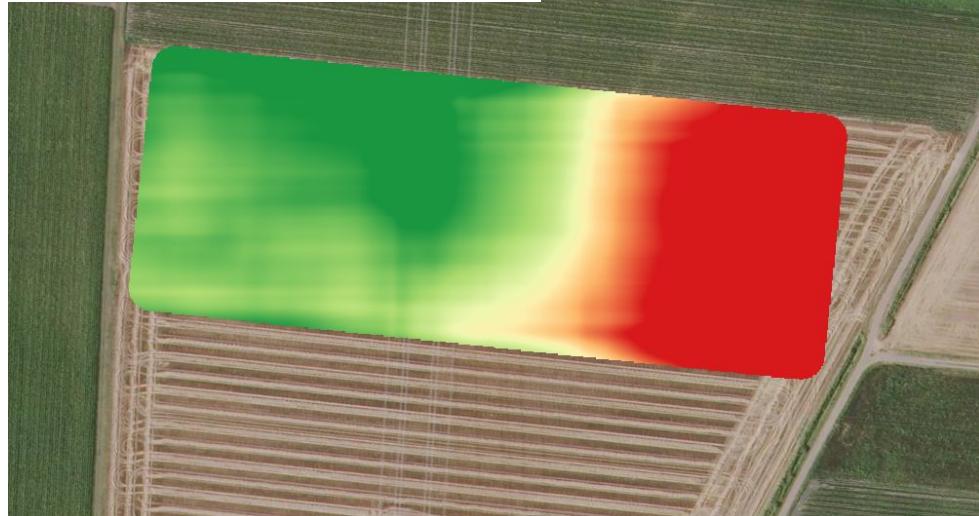
DSM



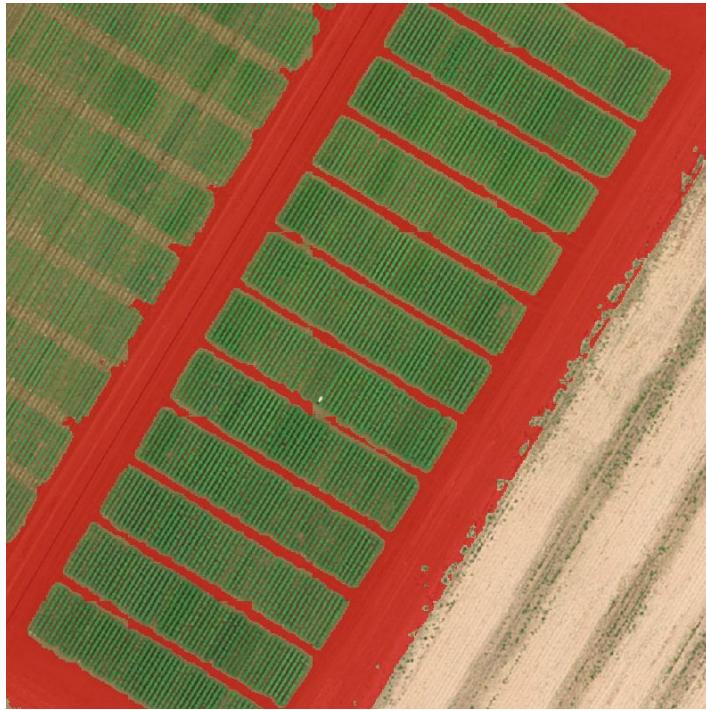
DTM



■ Digital Surface Model  
■ Digital Terrain Model

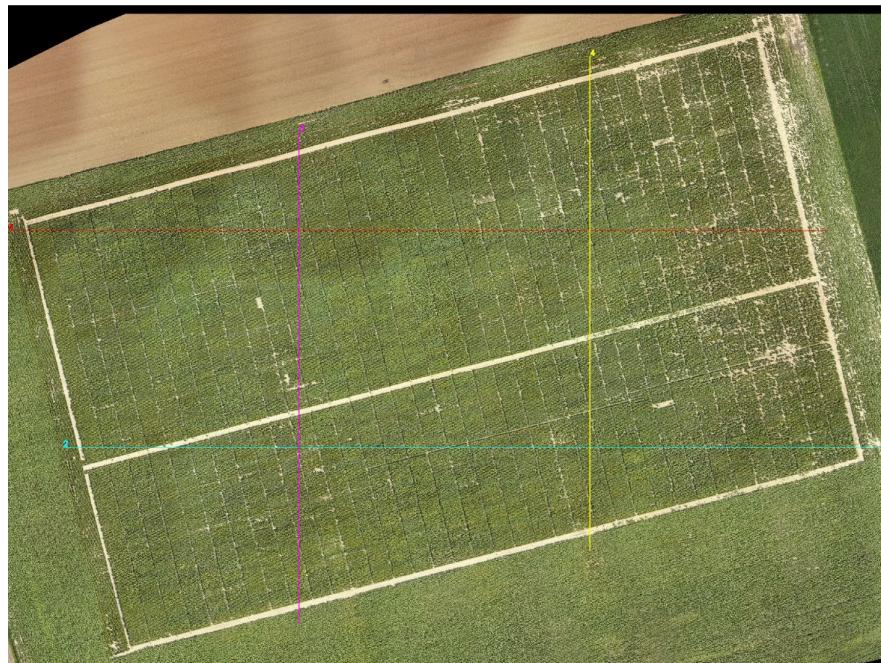
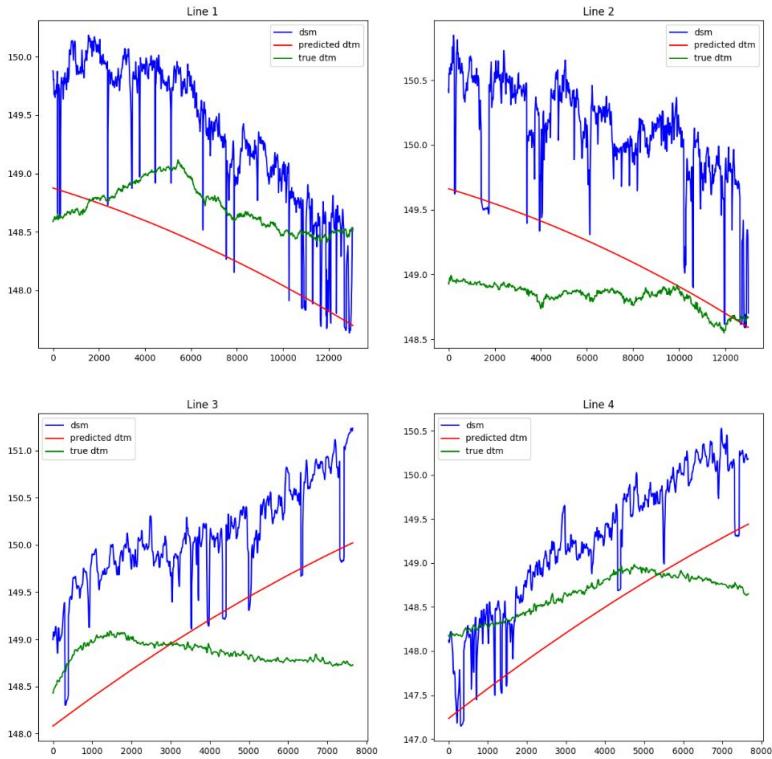


## Ground detection



- LinkNet
  - Semantic segmentation on ground
- IoU is ~70% but it is not a problem

## Results



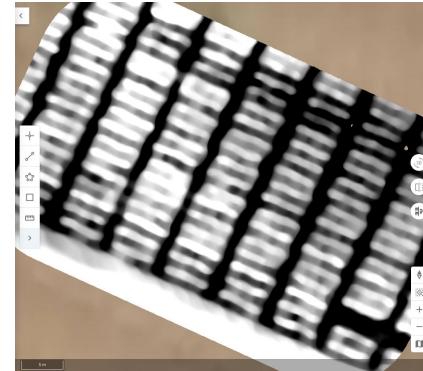
## The big picture



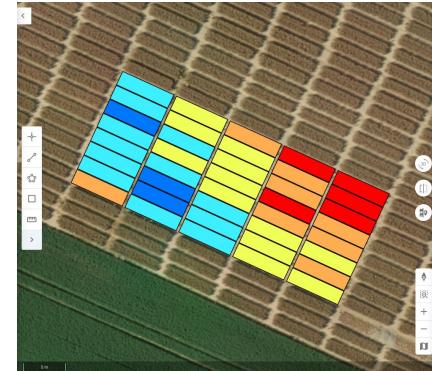
Capture: Raw Data



AI: Semantic Data



Digitalization: height model



Analytics: KPI

Our solution involves a multistep approach that transform progressively the raw data and each time provides new insights for the customer.

# CASE STUDY 2

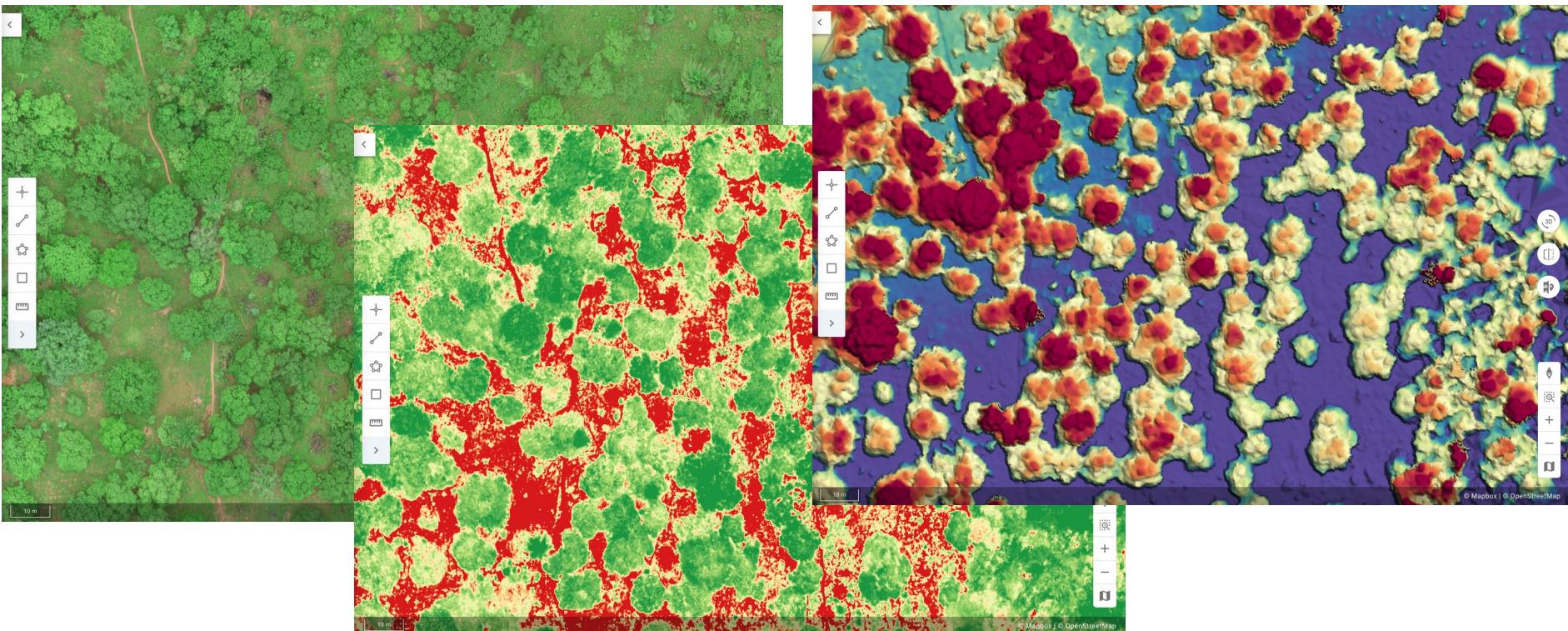
## Tree Counting

## Forestry inventory

- Count how many trees there are in a certain area
- Define position, surface, height



## Data available

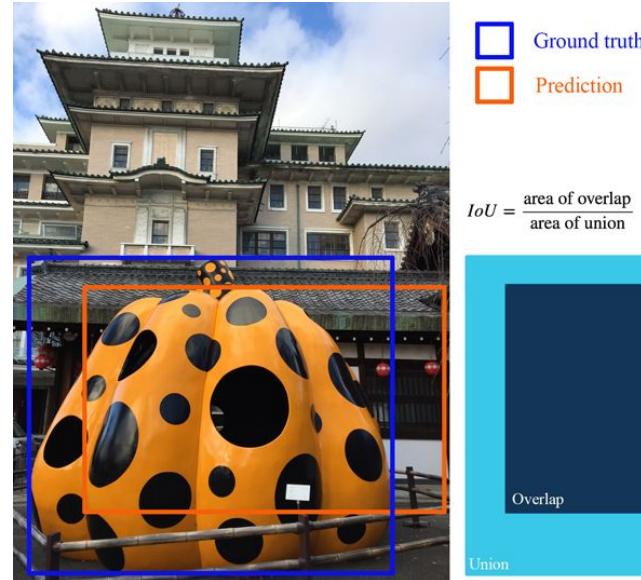


## Bounding box labels



## Supervised model

- Yolo
  - single stage detector
  - Fast and small
- Faster-RCNN
  - Two stage detector
  - Big and slow
- Uses multiple bands as input
- Metrics used:
  - mAP @ IoU = 50%
  - number of trees



■ Ground truth  
■ Prediction

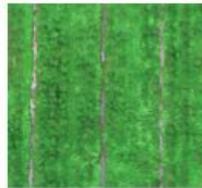
$$IoU = \frac{\text{area of overlap}}{\text{area of union}}$$



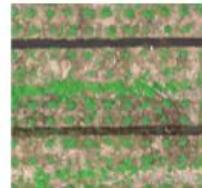
## Some qualitative results



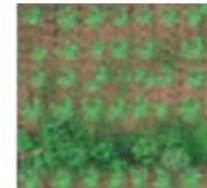
## Domain adaptation



FN161



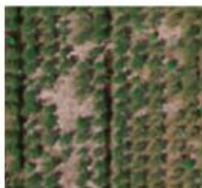
FN181



FN221



FN253



FN168\_North



FN53



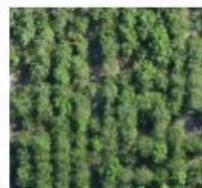
FN181



FN208



FN85



FN310

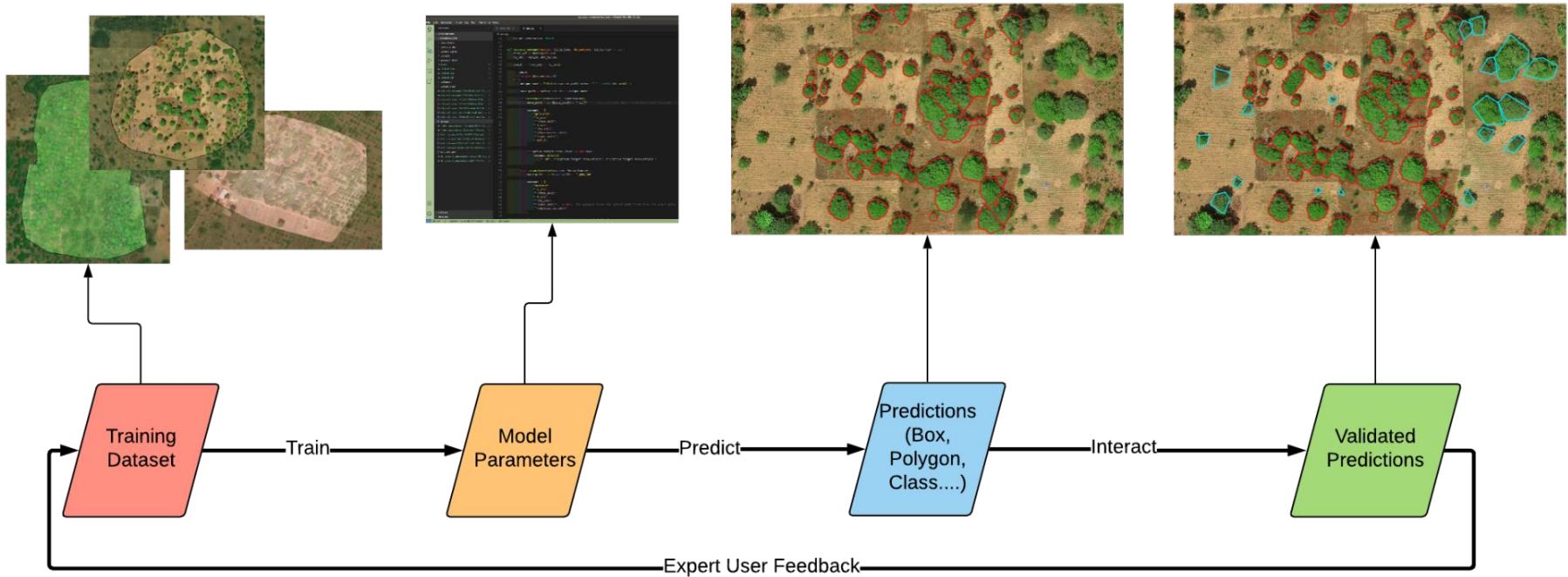


FN168\_South



FN86

## Interactive learning



# Conclusions

## Key take away

- One big mistake that everyone make is to think that humans are always 100% accurate.  
Measure, don't guess !
- Fit the best **sensor - vector- algorithm - data** set with the problem you are trying to solve.  
The 4 are interconnected and not always easily interchangeable
- Data is more important than models (talk [\*\*From Model-centric to Data-centric AI\*\*](#))
- The problem is more important than frameworks

## Alteia Team



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**Hacene  
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engineer



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**Aurélien  
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Computer vision  
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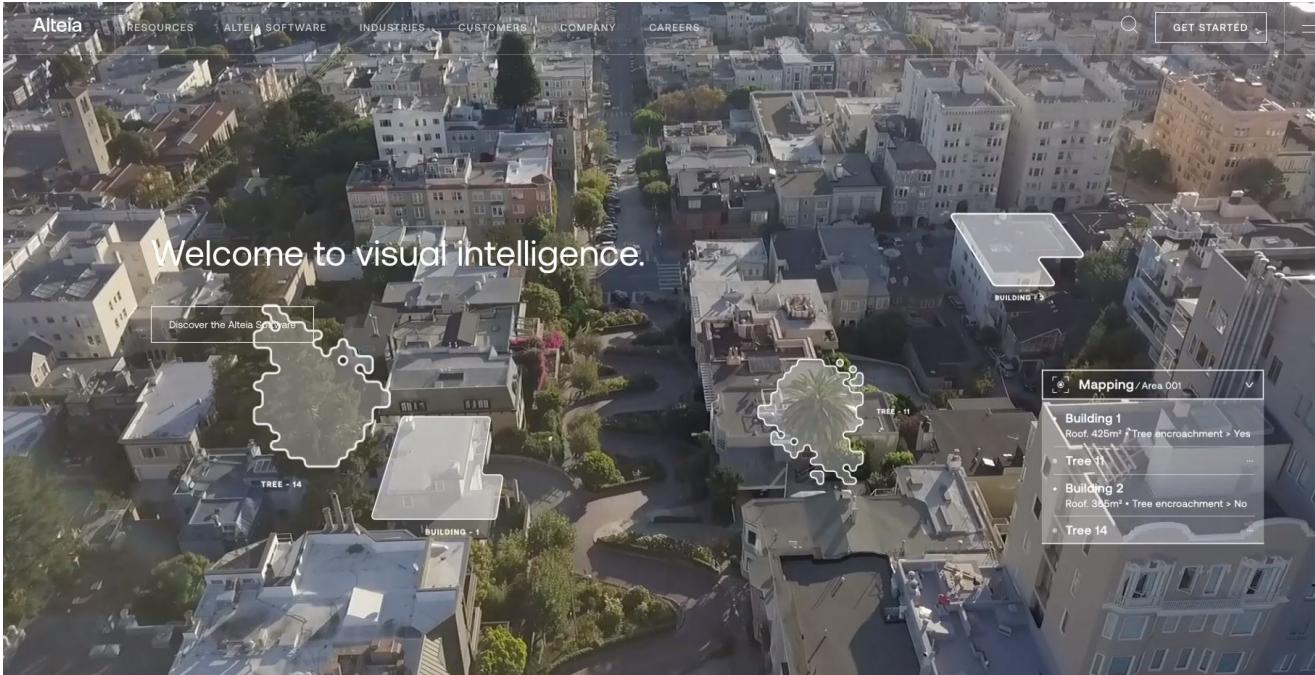


**Nicola  
Luminari**  
Head of Data  
Science



**Alexis Janson**  
Applied Product Engineer  
for Agriculture & Forestry

## Data Scientist Wanted

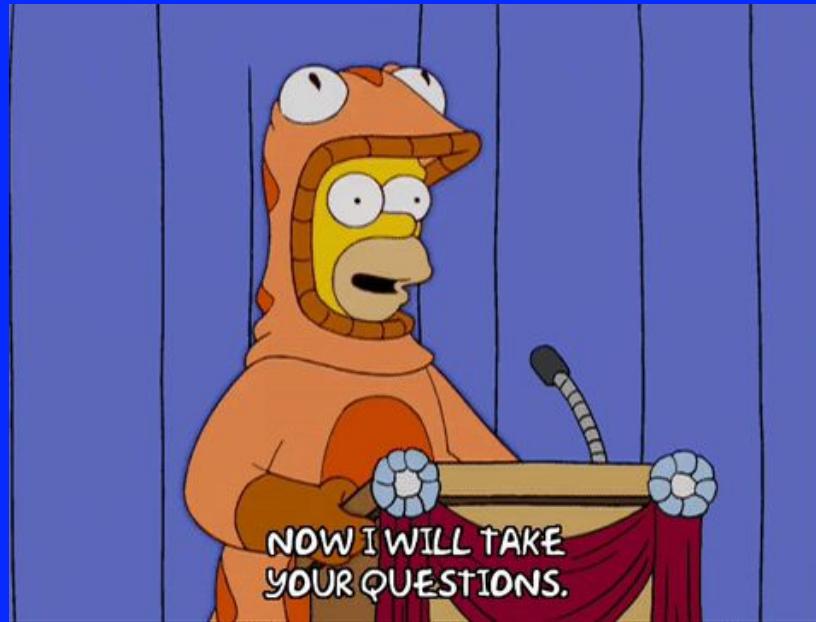


[Join Our TEAM !](#)

## Bibliography

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- [https://www.researchgate.net/publication/227304926 Kriging interpolation methods in geostatistics and DACE model](https://www.researchgate.net/publication/227304926)
- <https://github.com/davemlz/awesome-spectral-indices>
- <https://www.kaggle.com/c/global-wheat-detection>
- <https://alicevision.org/#photogrammetry>



NOW I WILL TAKE  
YOUR QUESTIONS.



**THANK YOU!**

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