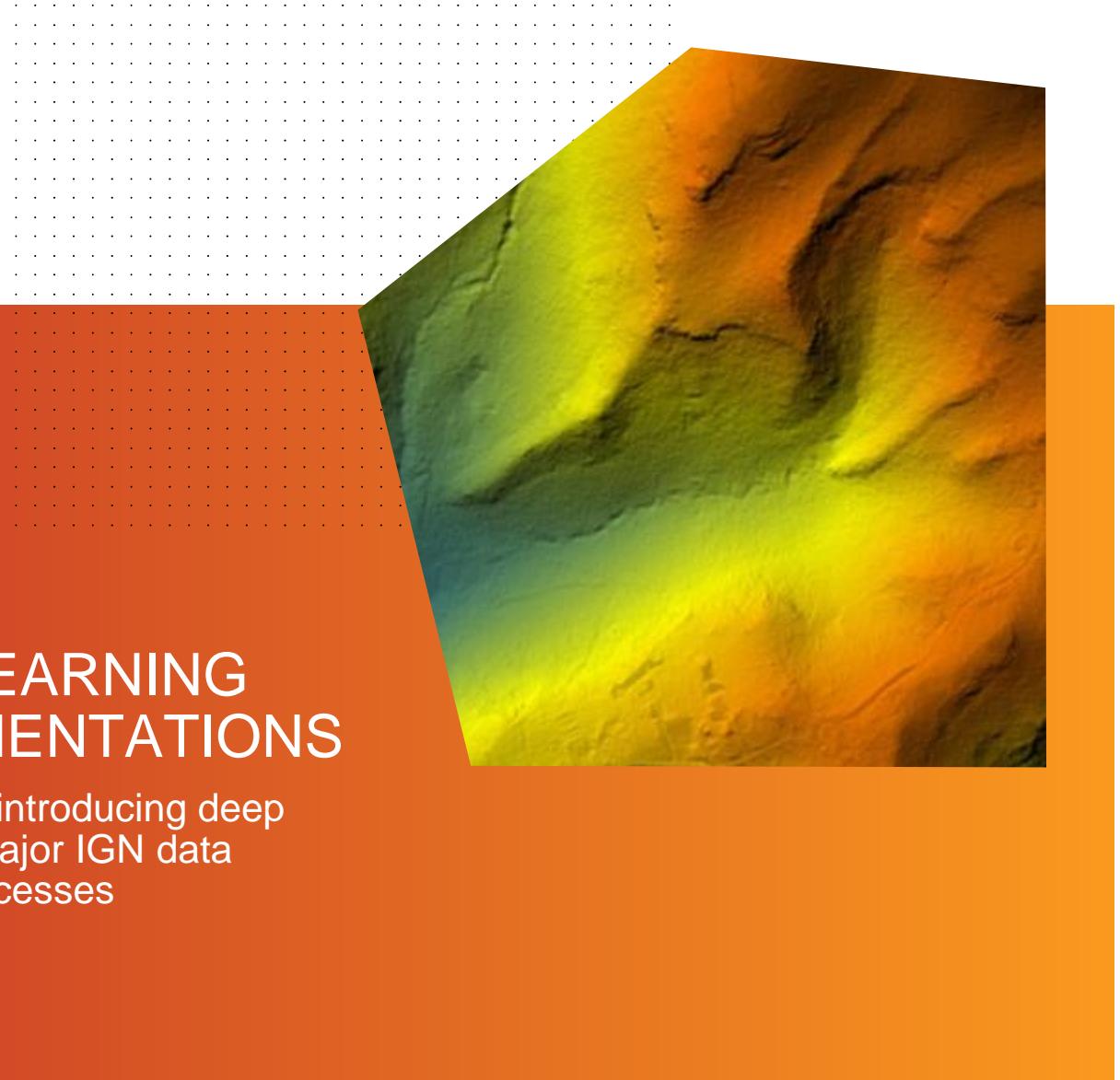




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DEEP LEARNING EXPERIMENTATIONS

First steps in introducing deep
learning into major IGN data
production processes



Content

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2. Deep Learning 101
3. Insights from a year of experiments
4. Zoom on timber : data generation
5. Summary and perspectives

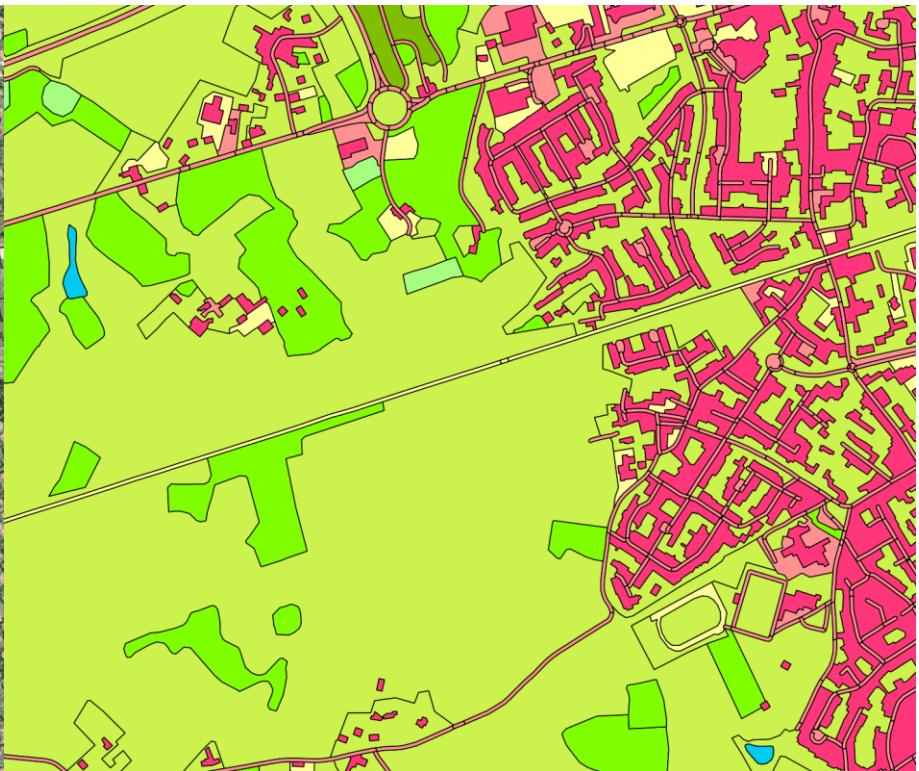


Illustrative example

Land artificialisation monitoring based on
OCS GE, a land cover/usage map



ORTHO HR®



OCS GE

Orthophotos (25cm/pix), Land Cover (14 fields)

ÉVOLUTION DU PROCESSUS DE PRODUCTION ET DE MISE À JOUR DE L'OCS GE

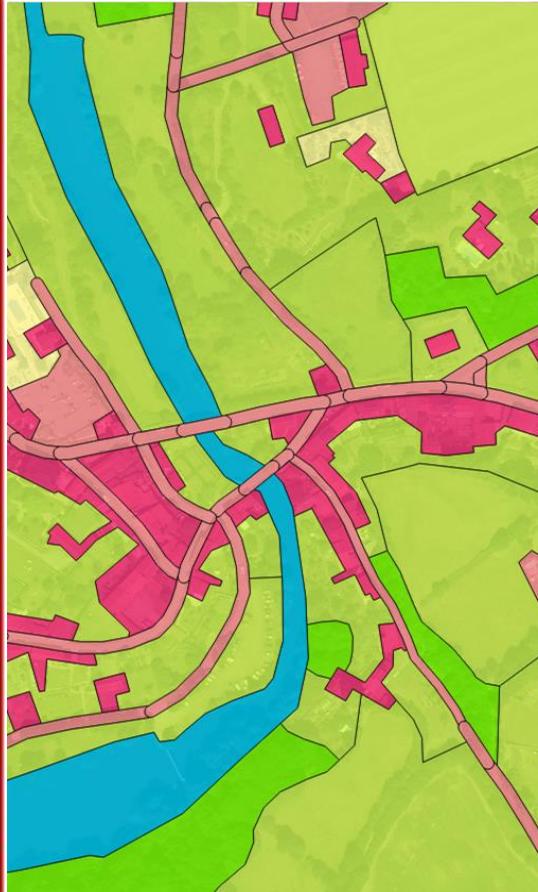
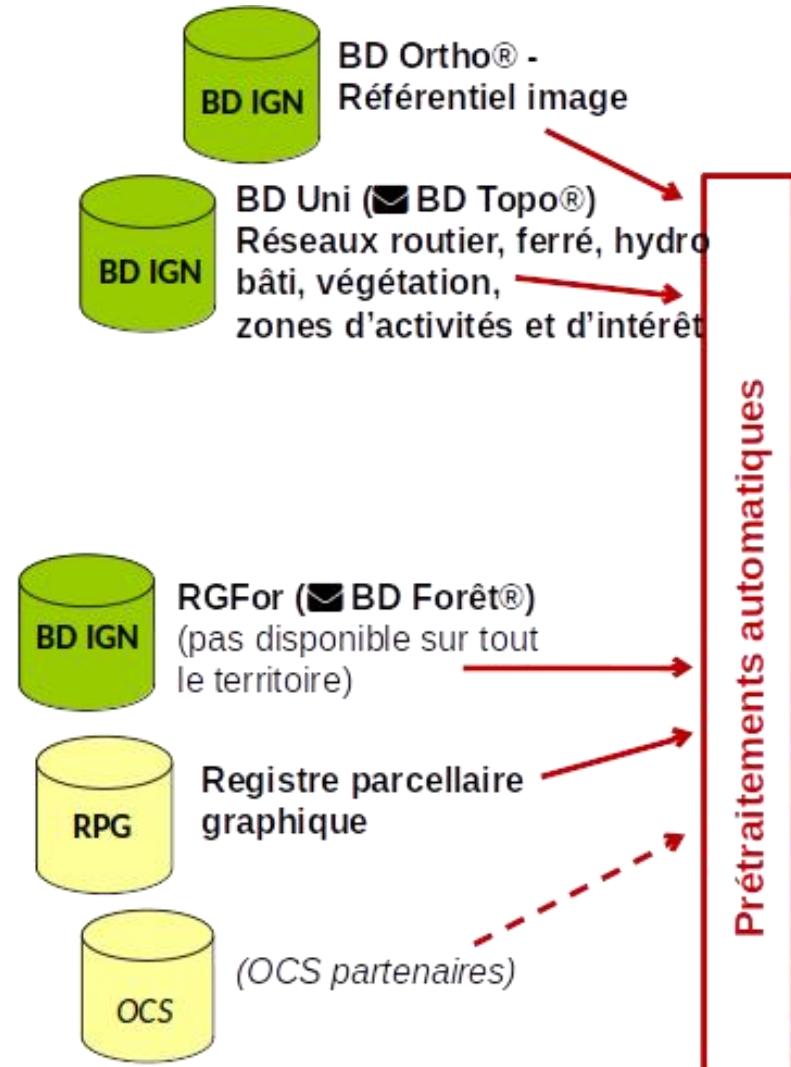


Photo-interprétation

Current process

Current project using OCS GE

OCS GE Qualities / Defaults

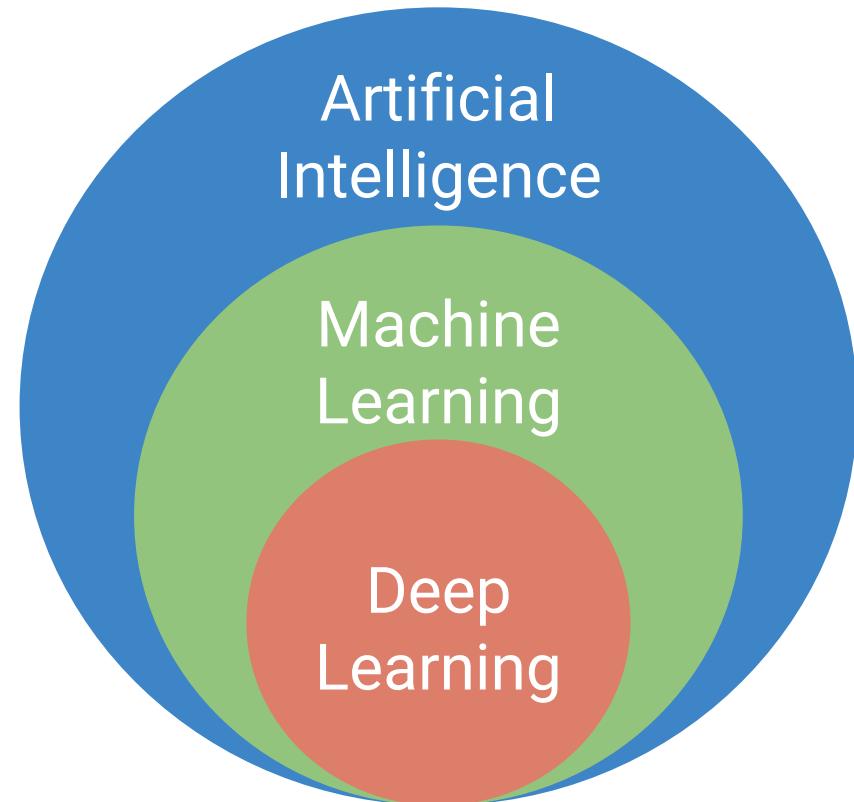
- better definition than other land cover products
 - Corine Land Cover : 25 ha
 - OCS GE : 200 up to 2500 m²
- sufficiently detailed to build land artificialisation indicators
- updated every 3 years (French territory)
- complex to build
 - lots of external sources
 - lots of manual processes
 - change detection and updates processes are not yet fully defined

Expected improvements

- **minimizing** the number of external resources
- **limiting** manual work
- **shorter** chain of process
- computation launched **as soon as** the orthophotos are produced
- **no need to wait** for updated databases
- **stability** of the detection year after year

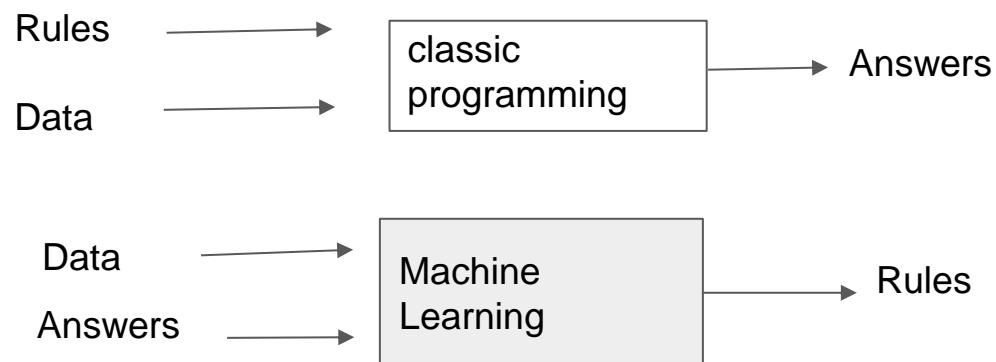


Deep Learning 101



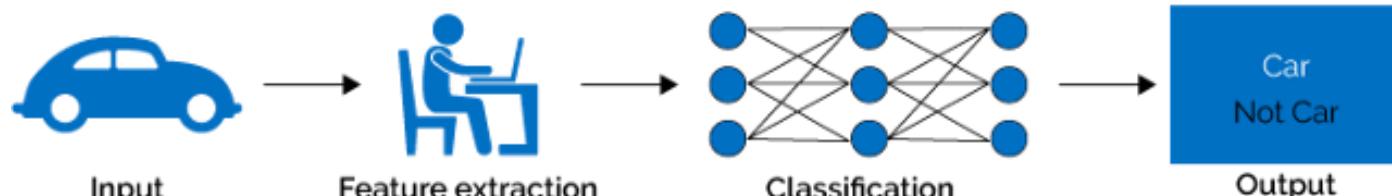
AI domains :

- planification
- natural language processing
- optimisation
- expert systems
- games
- ...

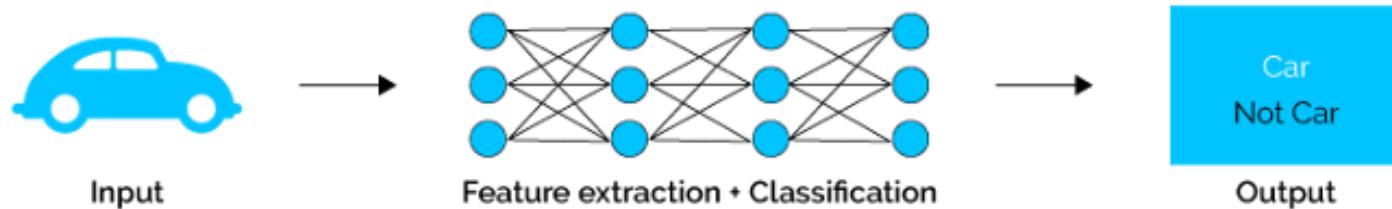


Machine Learning vs Deep Learning

Machine Learning

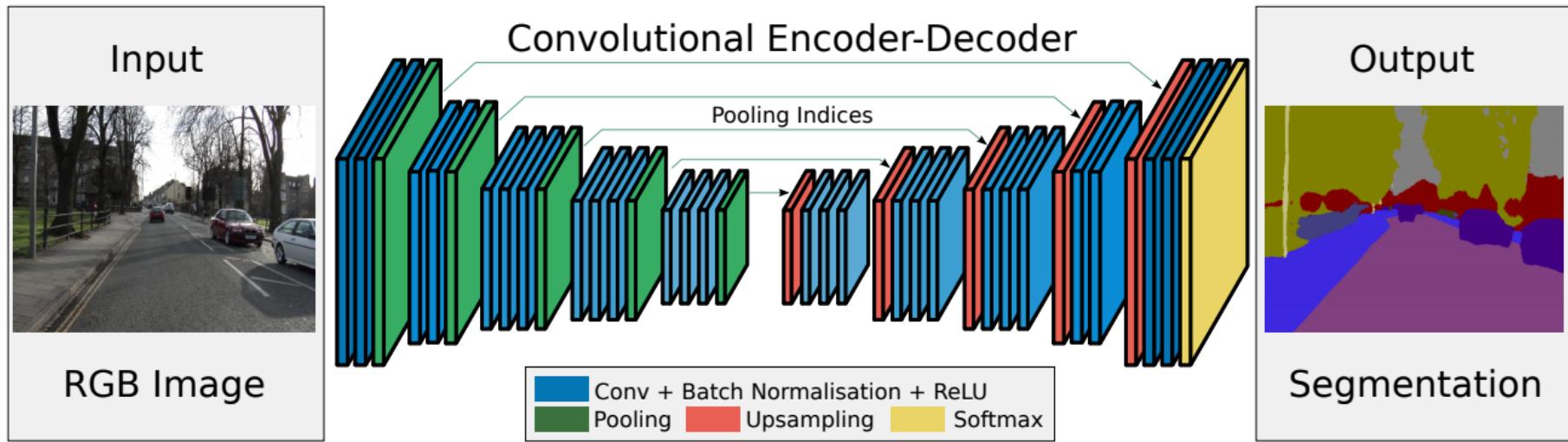


Deep Learning



- Less manual work
- Proven better than ML for image interpretation

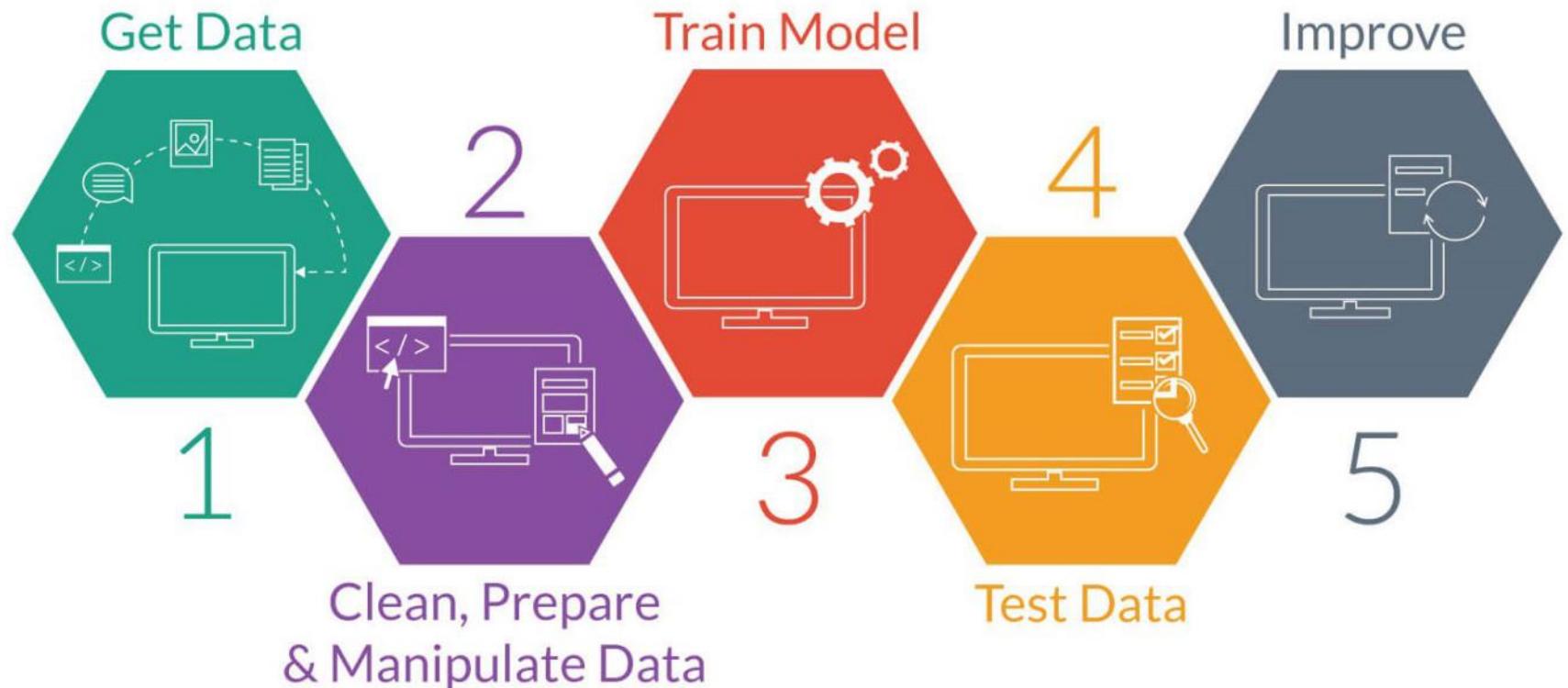
At the core of deep learning : a model



Model architectures :

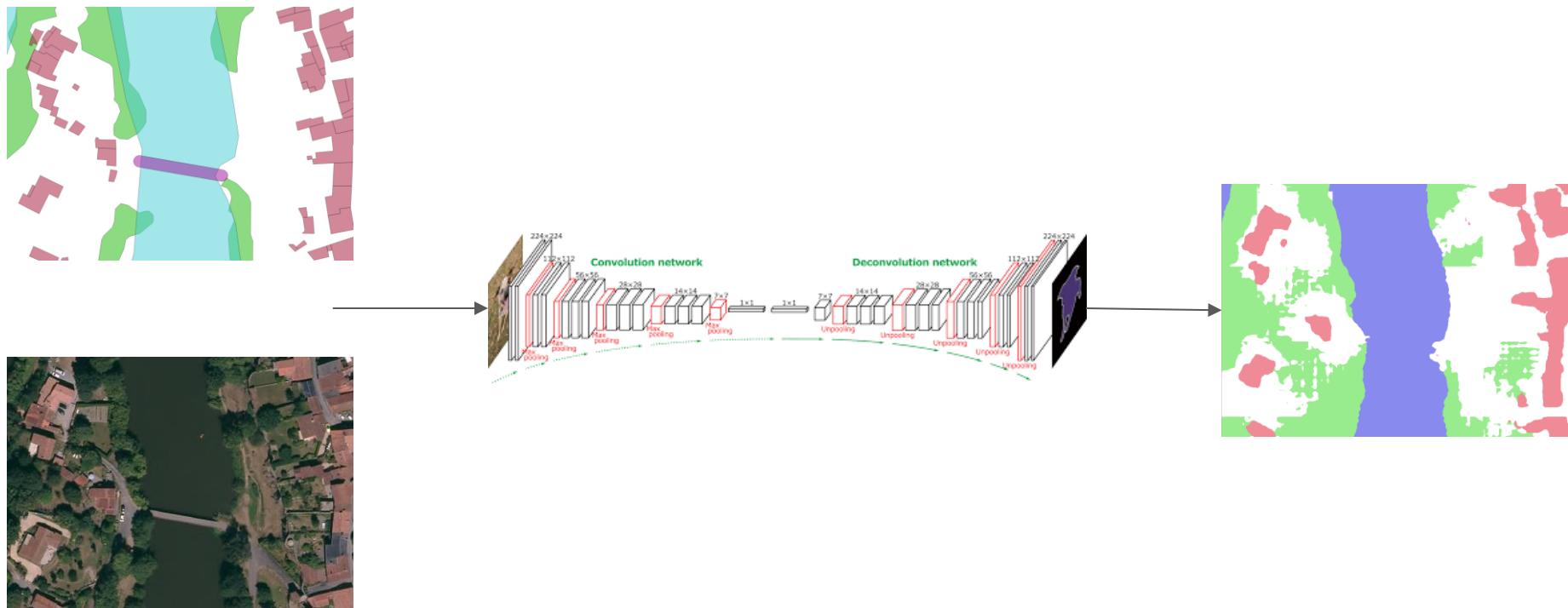
- **Convolutional Neural Network** (image, video)
 - image classification
 - object detection
 - **semantic segmentation**
- Recurrent Neural Network (temporal data, text, sound)
- Generative Adversarial Network (image, text, sound)
- ...

Deep Learning workflow (before deployment in production)

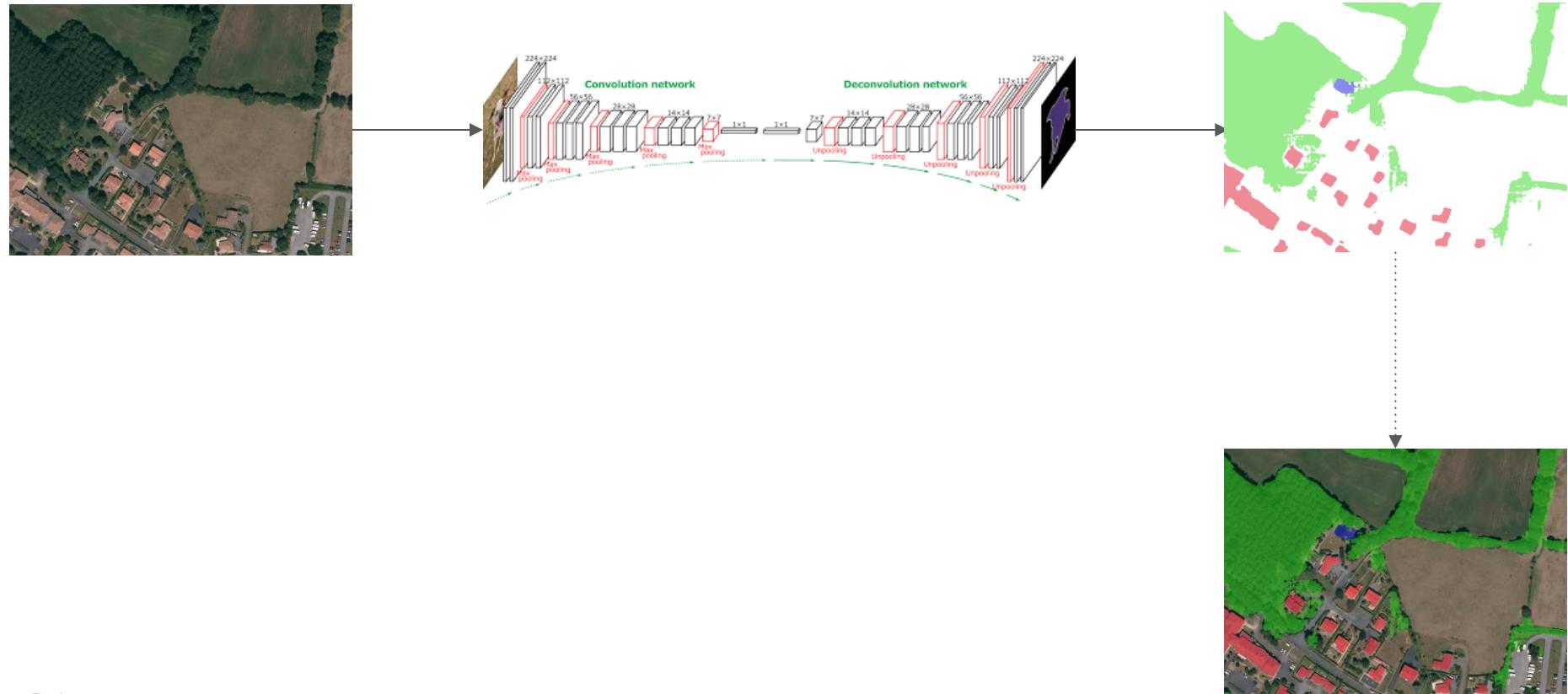


Once the models are stabilized and adequate
→ Integration into a production pipeline

Deep Learning : learning phase



Deep Learning : detection phase





Insights from a year of experiments

Organization - timeline

Summer 2018 (1 month)

Hackaton (2 days)

Termos v1
(4 month)

- Gathering of **raw data** (aerial & satellite, vectorized data)
- **Infrastructure** setup
- **Data set** generation
- **Basic working codes** for several objectives (forest id, detection of change in buildings, building segmentation)

- **Presentation**
- **Modification** of the proposed codes
- New and better datasets
- **Evaluation** (metrics, visualisation)
- **Improvement** proposals

- More **diverse datasets**
- New **models**
- Generation of **metrics** to rank the results
- **Bigger** territories investigated
- **Multiclass** detection
- ...

Termos v2 and beyond

- **Training** of additional personnel
- New **objects** of interest
- **Target** : OCS GE

Organization - workforce and infrastructure

Developers and data specialists

- Deep Learning
- Satellite data
- BDUni
- RGFor
- OCS GE
- ...

Computation power

- 6 upgraded computers with GPU (8 and 11 GB) and 32 GB of RAM
- 6 workstations for code development, data treatment, data visualisation, etc.

Lots of meetings with :

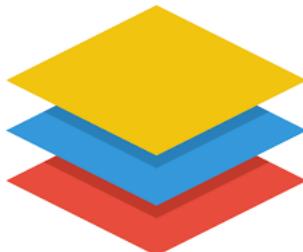
- researchers
- potential end users
- decision makers
- software developers
- ...

Mode of operation



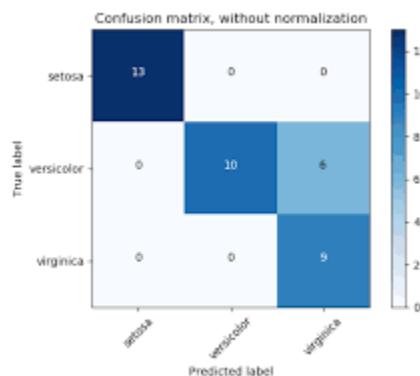
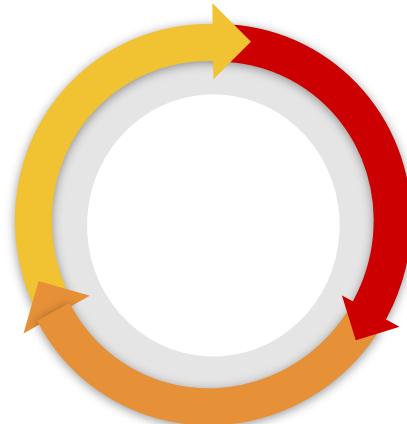
Raw data sources :

- BD ORTHO
- BD ALTI
- BD UNI
- RG For
- SPOT
- ...



Training datasets :

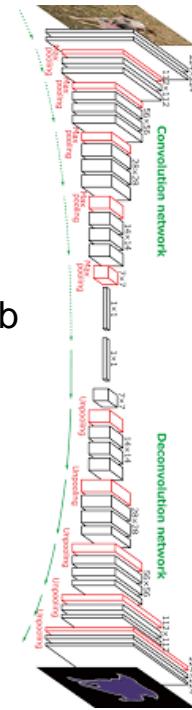
- several zones in France
- several objet type



Evaluation :

- qualitative
- quantitative

Model Comparaisons Analysis of detection failures



Ground truth from BDUNI/RGFor



Termos v1 Deep Learning output





Zoom on timber

or how to relatively quickly build a ground truth data

What is in the database (RGFor)



What we need : exhaustivity



Methodology using Orfeo toolbox

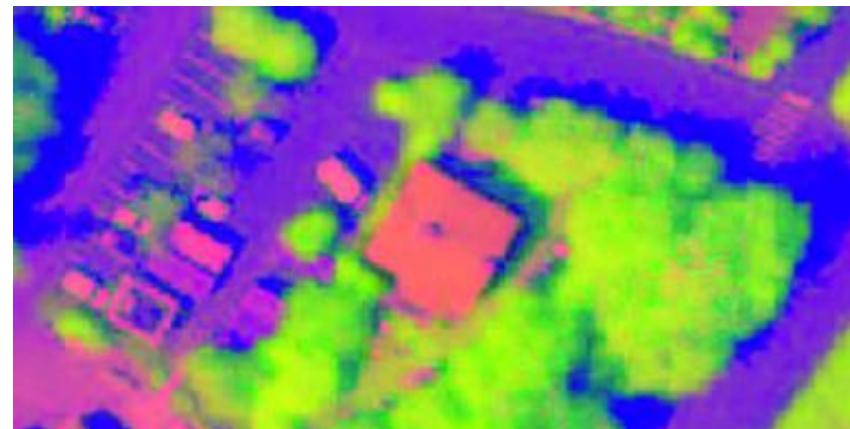
Apply a geometric segmentation at the right level of detail

→ no shadows, separation between trees and grass



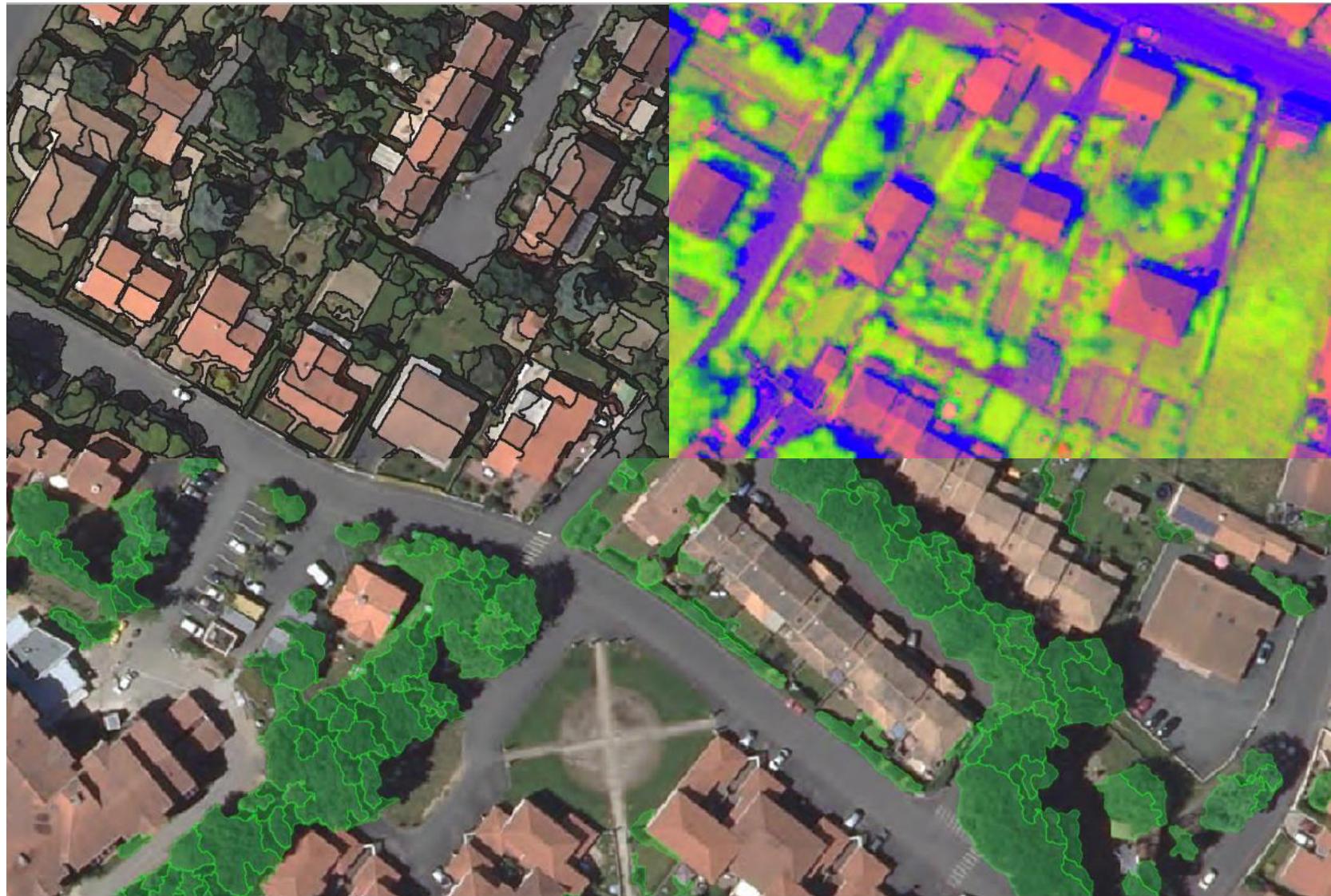
Use Machine Learning

- Manually define features (ndvi, bi2, ...)
- Apply Random Forest algorithm to distinguish timber from the rest



Orfeo ToolBox: Open Source processing of remote sensing images (<https://www.orfeo-toolbox.org/>)





Geometric segmentation and feature representation (top)

Machine learning output (bottom)



Summary and perspectives

Up to date results : better timber, work in progress on asphalt



What is a “good data” ?

Nomenclature

- IGN: complex (minimum surface, mixed component)
- DL: what you see is what you learn / get

Visual rendering

- IGN: vector
- DL: raster

Semantic

- IGN: usage or cover
- DL: mostly cover

IGN strength

We are currently in the process of producing a very accurate and clean dataset for **urban areas** (with 7 general classes and several subclasses) in France (10 “departements” from 2015 up to 2018)

→ datasets are the most valuable asset in the Deep Learning community (famous dataset : ImageNet, Mapillary, ...)

We have access to more data, particularly height information, close infrared images

→ allows us to limit the size of dataset (<< 1M samples) but we can't profit from pre-trained models

Perspectives

- Change detection and alert system
- Accelerate the update of main IGN databases (BDUni, RGFor, ...)
- Fault detections during the transformation from PVA to orthos photos
- Real time Mobile Mapping
- ...



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