From supervised to unsupervised deep learning for automatic detection of marine megafauna



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# Semmacape project

Use high resolution images and deep learning methods to map marine megafauna





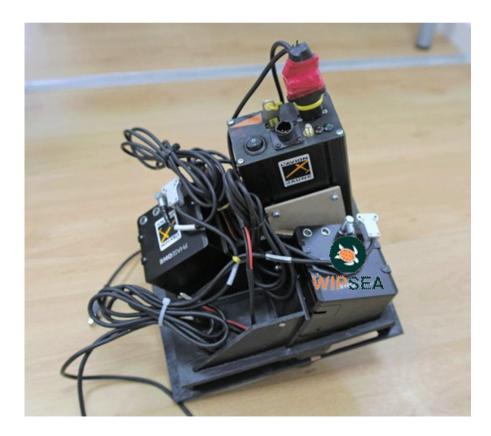
#### Issue

- More and more organizations need to census megafauna
  - Offshore wind farms
  - Marine protected areas
- Problem: heavy to deploy, expensive, lacks accuracy



## Solution

- High resolution aerial imagery
  - Shot with WIPSEA and/or partners cameras
  - Verifiable by « picture proof »





### Solution

 Megafauna automatic detection and classification software solution (Deep Learning) to manage Big Data



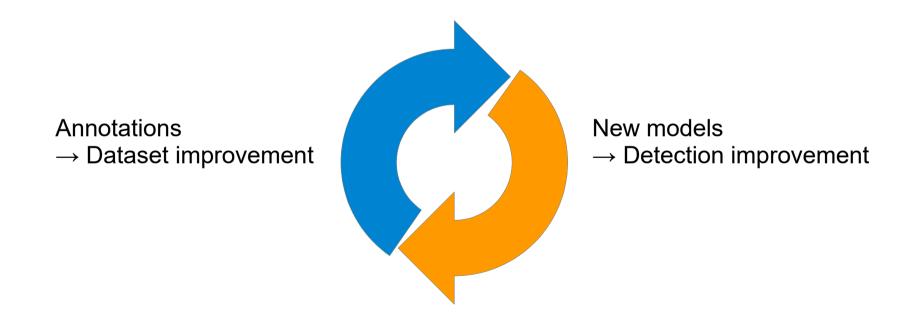
# Solution

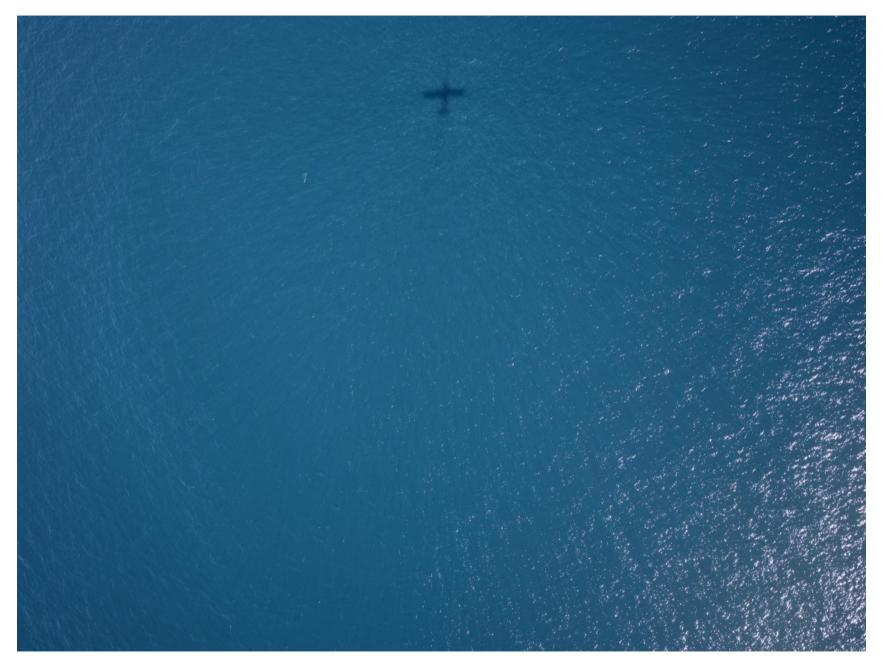
- > 100 000 high resolution photos at 180m altitude already shot (~3000 km<sup>2</sup>)
  - Goal : develop robust solutions for marine megafauna aerial surveys and compare detections with onboard visual observations
- R&D collaborative project funded by Agence De l'Environnement et de la Maîtrise de l'Energie
- Partners :
  - IRISA, Office Français de la Biodiversité
  - France Energies Marines, IFREMER



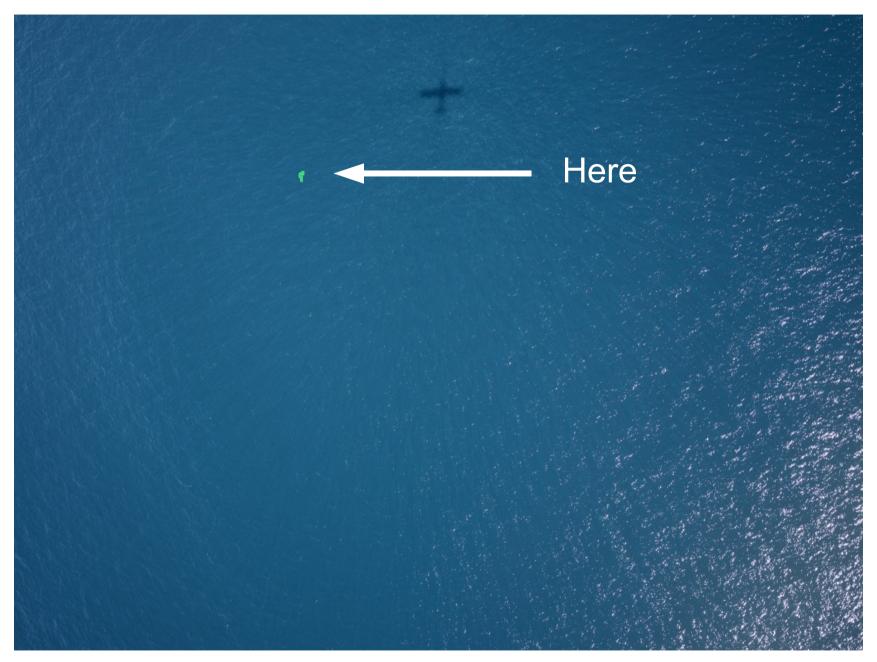
## Dataset and detection models in a supervised context

- Iterative dataset/model building
  - Megafauna in the mass of photos are, at first, like « a needle in a haystack »

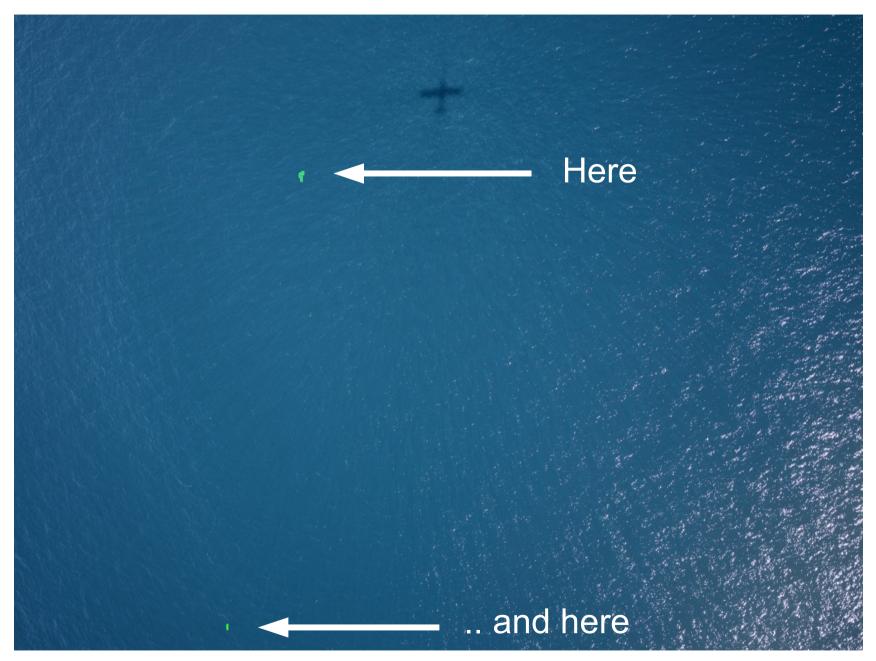




A needle in a haystack : where are the dolphins ?



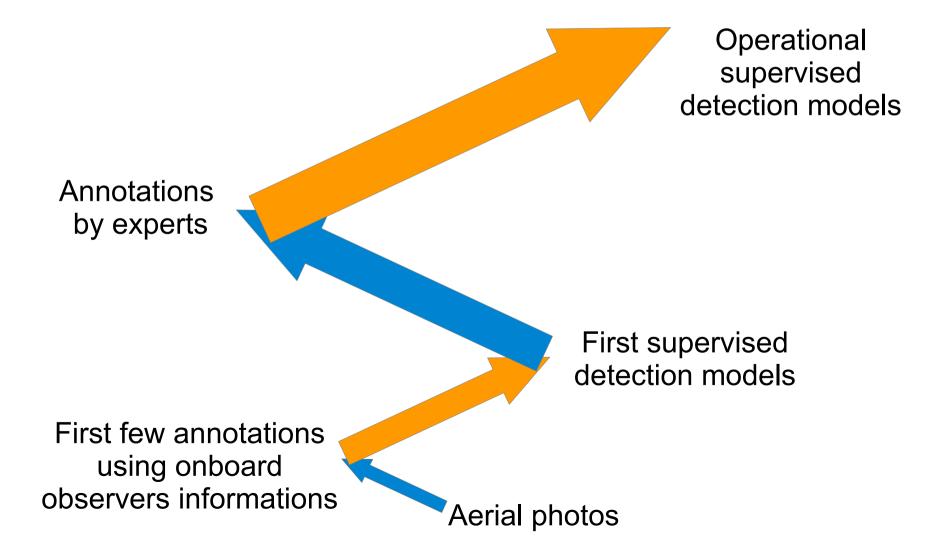
A needle in a haystack : where are the dolphins ?



A needle in a haystack : where are the dolphins ?

## Dataset and detection models in a supervised context

Iterative dataset/model building



### Up-to-date dataset

- 6248 photos (~180km<sup>2</sup>) validated by an expert
- Covers spring, summer and autumn
- 910 marine mammals inventoried
- 18443 birds
- 4229 big fishes and sharks

- Test procedure
  - Operational test set ≠ test on objects only
  - As opposed to many dataset, objects are needles in a haystack
  - Tests apply on hundreds of images, representing tens of square kilometers
  - Aerial images of the sea can show many different shapes likely to trick the detector
- As 'emptiness' grows in the test set, the detector is more likely to be tricked (false positives issue)

Intersection

over

union @ 0,5

Current operational performances

**Dolphins**, **Birds Birds Big fishes**, **Jellyfishes** flying landed sharks Average 0,12\* - 0,77 0,49 - 0,56 0,37 - 0,68 0,34 - 0,82 precision 0,49 - 0,77 0,57 - 0,68 0,46 - 0,70 0,19\* - 0,75 Max F1 score

Tested on areas from ~20 km<sup>2</sup> to ~50 km<sup>2</sup> ~600 to ~1600 images Close to operational conditions

\* For now, score varies depending on the test set (season, species population, etc..)

 $\rightarrow$  Consolidation phase

- Current performances
  - Dolphins, big fishes, sharks

 $\rightarrow$  Fully/partly underwater, hard to detect

- Birds
  - $\rightarrow$  Different species with very different visual features
    - Size, shapes, contrasts

- Conclusion
  - Supervised detection is already effective for the census of megafauna
    - Major help for expert image analysis with dedicated tools
- Work in progress
  - Model improvement using knowledge on data and state of the art methods
  - Detect more specific classes

• What if there are no available annotations ?