

Introduction

Motivations

- Large demand for algorithms that can help domain experts in the analysis of huge image datasets for ecological studies
- Vital importance of the study of cetaceans
- Recent publication of an algorithm for the automated photoidentification of Risso's dolphins [1]

Contribution

Development of a Matlab routine able to automatically detect dolphin dorsal fins in images, based on a new, efficient algorithm which combines domain analysis and deep learning [2].



Figure 1: Visual representation of the problem

Data

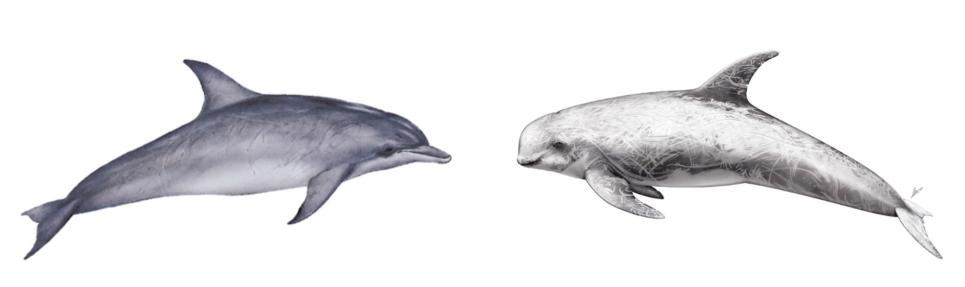
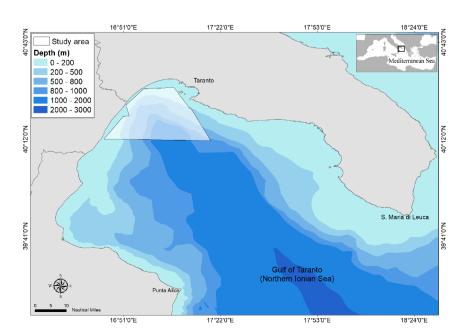


Figure 2: (Left) Tursiops truncatus, (Right) Grampus griseus (Risso's dolphin)

Tursiops and Risso's dolphins images collected by two private reasearch associations:

- $\sim 10,000$ pictures taken in the Gulf of Taranto (Northern Ionian Sea) between 2013 and 2018
- $\sim 14,000$ pictures taken near Pico island (Atlantic Ocean) in 2018



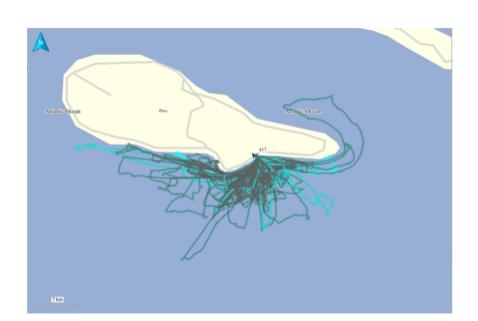


Figure 3: Study areas

Combined Color Semantics and Deep Learning for the Automatic Detection of Dolphin Dorsal Fins

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Methods

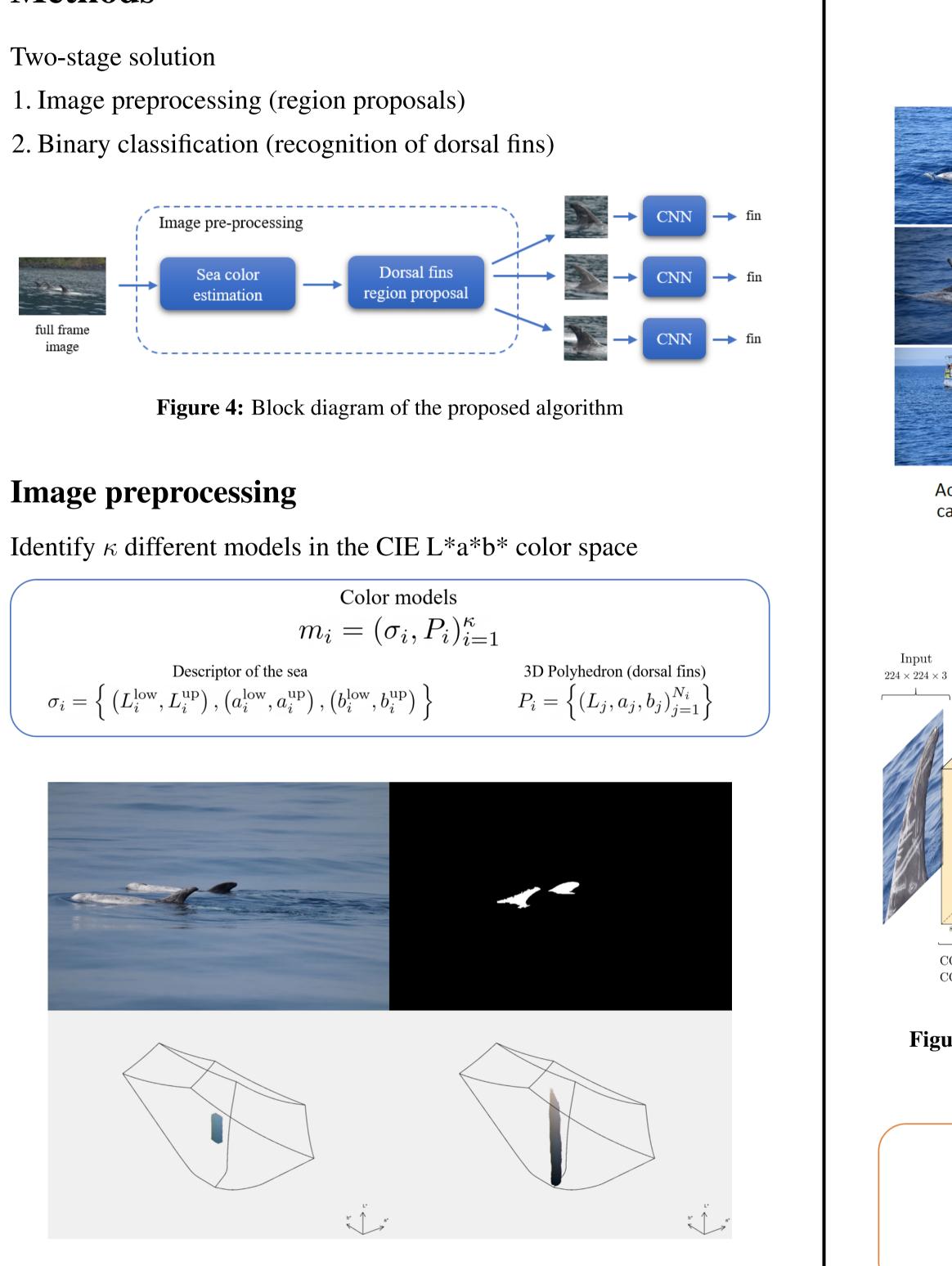


Figure 5: Example of segmentation

Preprocessing of image *I*:

1. Model assignment (sea color estimation):

$$i^{\star} = \arg \max_{i=1}^{\kappa} \sum_{j,k} \mathbb{1} \left\{ I_{jk} \in \operatorname{sea}(\sigma_i) \right\}$$

2. Region proposals (binary mask with $P_{i^{\star}}$, morphological operations, size refinement, aspect ratio filter)

Binary classification

Create a dataset with the image preprocessing algorithm and train a convolutional neural network (CNN) to recognize dorsal fins



Results

Experiment 1. Creation of $\kappa = 5$ models by applying a semiautomated iterative procedure on a small subset of images ($\sim 2,000$). Comparison of the results to an alternative approach based on Otsu's method.

Experiment 2. Training and validation of a custom CNN on 15, 228 cropped images from the Gulf of Taranto (random split 80-20%).

Experiment 3. Assessment of the performances of the CNN on 20, 888 cropped images from Pico island.

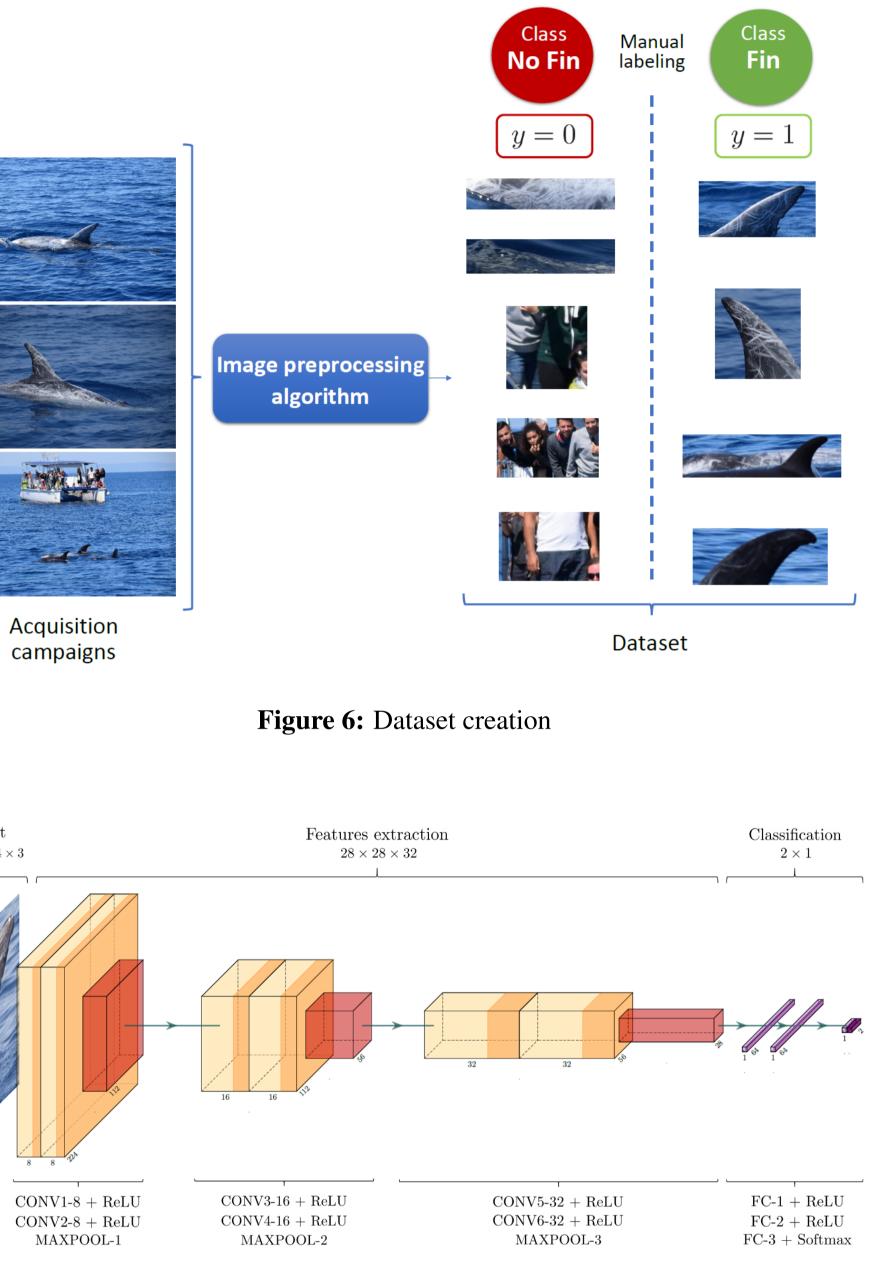
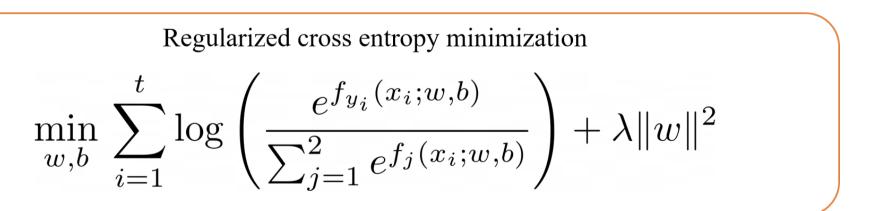
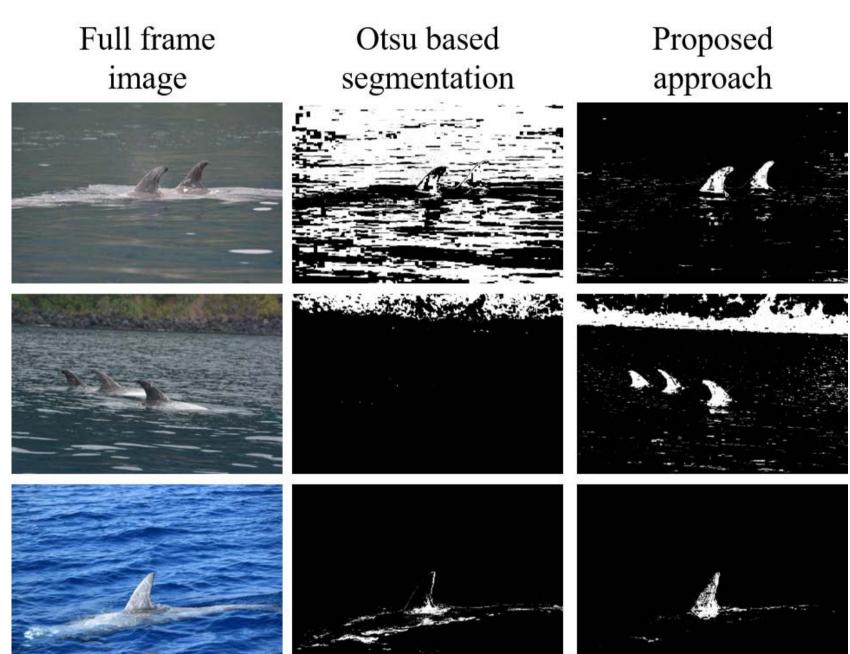


Figure 7: Custom convolutional neural network: 23 layers with 3×3 kernels







Validat Tes

Table 1: Performances of the classification. (\star) 5-fold cross validation strategy: mean values over 5 CNNs. (†) major voting: 4 CNNs out of 5. Training time single CNN: ~3h 20 mins on Intel Core i5-6400T, 8 GB RAM, Nvidia GeForce 930M with 2 GB

Conclusions

- High flexibility in splitting the region proposal task from the classification task (introduction of domain knowledge)
- High performances with a resource efficient CNN
- Feasibility of the proposed approach to enable large scale ecological studies of dolphins

References

- 2020.

Acknowledgements

Figure 2: credits to animalbase.uni-goettingen.de, associaciocetacea.org. Figure 7: credits to Haris Iqbal.





Figure 8: Qualitative analysis of the proposed segmentation method

	Accuracy	Sensitivity	Specificity
ation ^(*)	99%	99%	99%
$st^{(\dagger)}$	92%	85%	95%

Future works. (i) Test in different operating conditions. (ii) Extend the methodology to biological investigations of different species.

[1] Maglietta R. et. al. Dolfin: an innovative digital platform for studying risso's dolphins in the northern ionian sea (north-eastern central mediterranean). Scientific reports, 8(1):1–11, 2018.

[2] Renò V. et. al. Combined color semantics and deep learning for the automatic detection of dolphin dorsal fins. *Electronics*, 9(5):758,