Zebrafish Skeleton Measurements using Image Analysis and Machine Learning Methods

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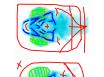
Context: Traditionally, biologists evaluate effects of experiments on zebrafish embryos through microscopic obsevations. However, visual inspection by experts is a limiting factor in large-scale studies. It brings us to consider supervised machine learning methods to automate the extraction of useful, quantitative, information from these images.

Problem: We focus on morphometric measuring of the cartilage skeleton. To this end, we address two subproblems: (i) quantifying the surface of the cartilage skeleton and (ii) detecting several points of interest in zebrafish images. We propose to tackle these two tasks by exploiting expert annotations with extremely randomized tree methods combined with extraction of subwindows within images [1,2].

Results: They are visually coherent despite the difficulty to assess our model in the absence of a complete ground truth annotation.

Segmentation of cartilage skeleton

Partial annotation into three classes: Eye, Skeleton and Others

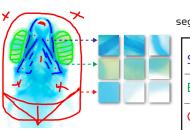


Detection of points of interest

Extraction of subwindows of size $\mathbf{w} \times \mathbf{h}$ centered on pixels in a radius \mathbf{r} around the points of interest and on pixels taken randomly (input), and tagged by the distances of their central pixels to the points of interest

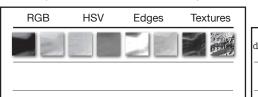
Multiple output regression model built to jointly detect multiple points of interests within images





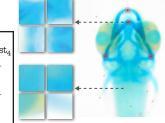


Skeleton Eye Other

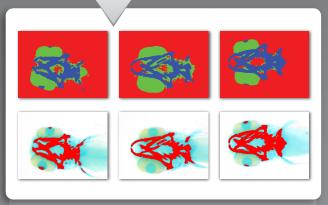


Input





edges (i.e. the gradient of the Sobel operator) and the texture (i.e. the histogram of local binary patterns [3]) of all the pixels in the subwindow.



The 12 remaining images were automatically annotated: we extracted subwindows centered on each pixel of every test images and the tree ensemble model predicted their class













Conclusions and perspectives

We present two methods able to automatically segment new, unseen, zebrafish images into a finite set of predefined classes, and to detect points of interest in it. Nevertheless, we worked on a small set of control images. The next step in this research is to apply these techniques on larger sets of new images of zebrafish presenting deformations to better assess their performances.

- Kimmel et al., The shaping of pharyngeal cartilages during early development of the zebrafish. Dev. Biol., vol. 203, pages 245-263, 1998.

 Dumont et al., Fast Multi-Class Image Annotation with Random Subwindows and Multiple Output Randomized Trees. Proc. VISAPP, 2009.

 Ojala et al., A Comparative Study of Texture Measures with Classification Based on Feature Distributions. Pattern Recognition, vol. 29, pages 51-59, 1996.
- Acknowledgements

