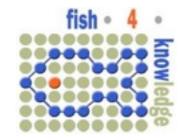


Experts, non-experts and automatic methods in crowdsourcing for wildlife image annotation

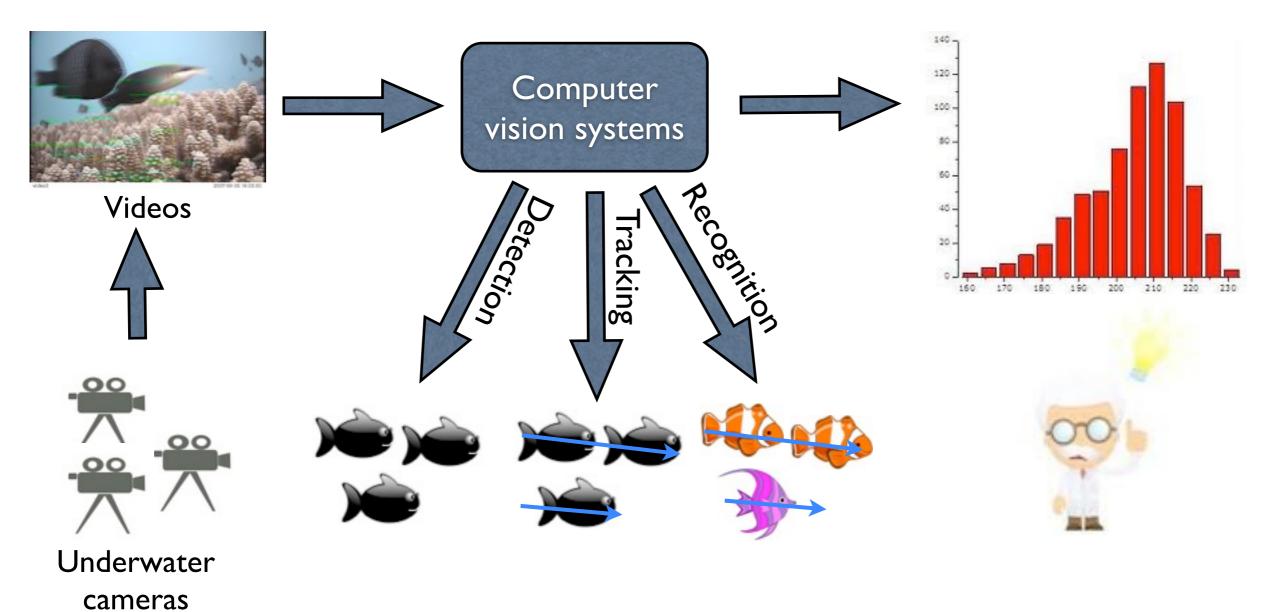
Jiyin He, CWI soHuman 2012



The big picture

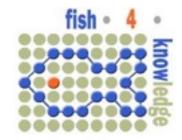


The Fish4Knowledge Project

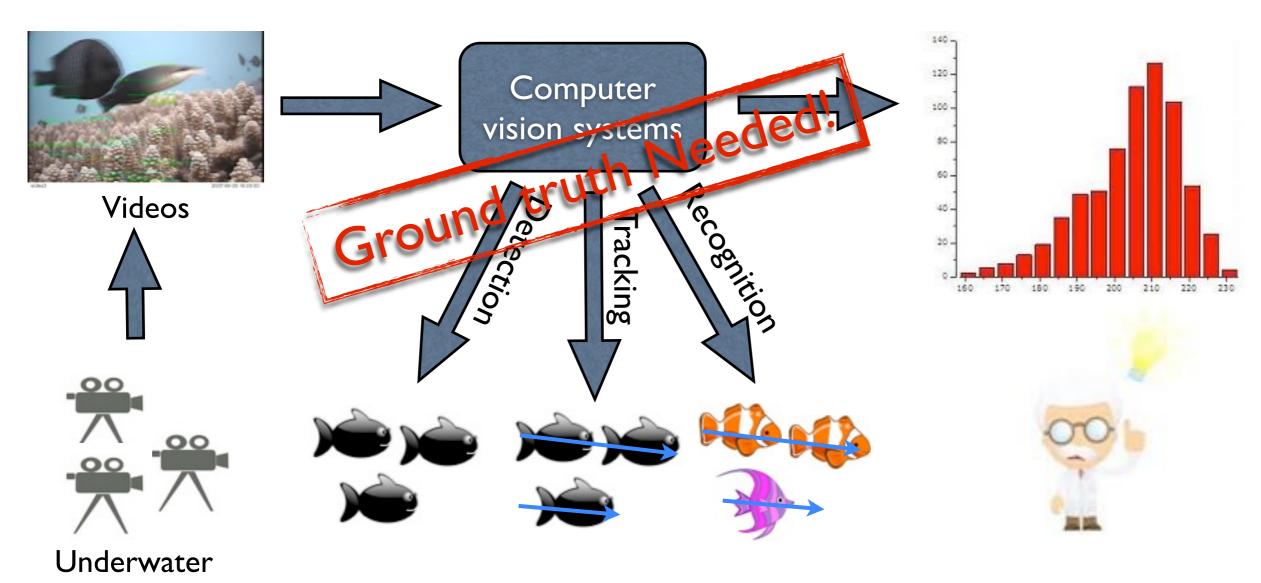




The big picture



The Fish4Knowledge Project



cameras

Fish recognition: a difficult problem

- Goal: given an image containing a fish, assign the species name of the fish to the image
- Large amount of images
- Expert knowledge needed
 - Non-experts often lack the knowledge needed to recognize a fish
 - Non-experts may not be able to map the common name of a fish to its scientific name
 - Even experts can have their expertise in different types of fish or fish in different areas
- Experts are expensive, rare resources

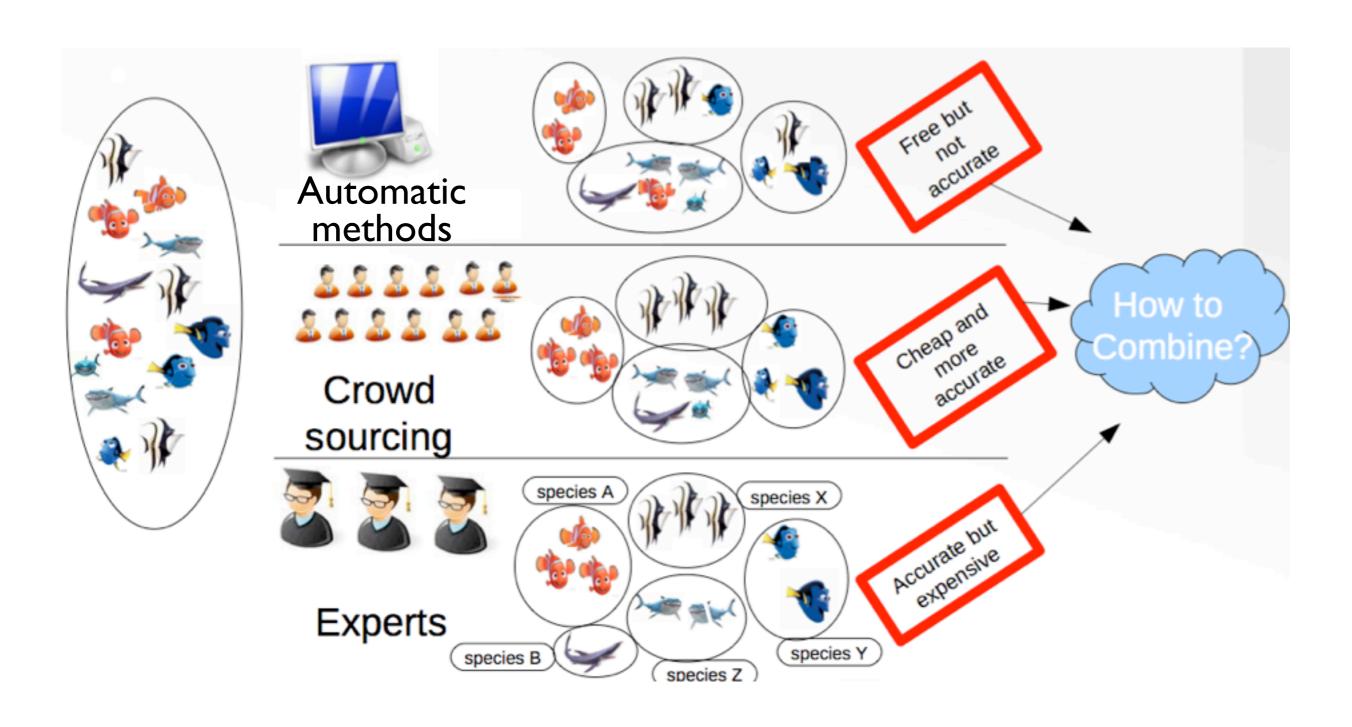


What can non-experts (not) do?

- Assumptions
 - Non-experts are not able to name fish species
 - But may be able to tell if two fish are visually similar
- Possible tasks
 - Manual clustering
 - Classification with textbook images as category labels



Automatic methods, crowds, and experts



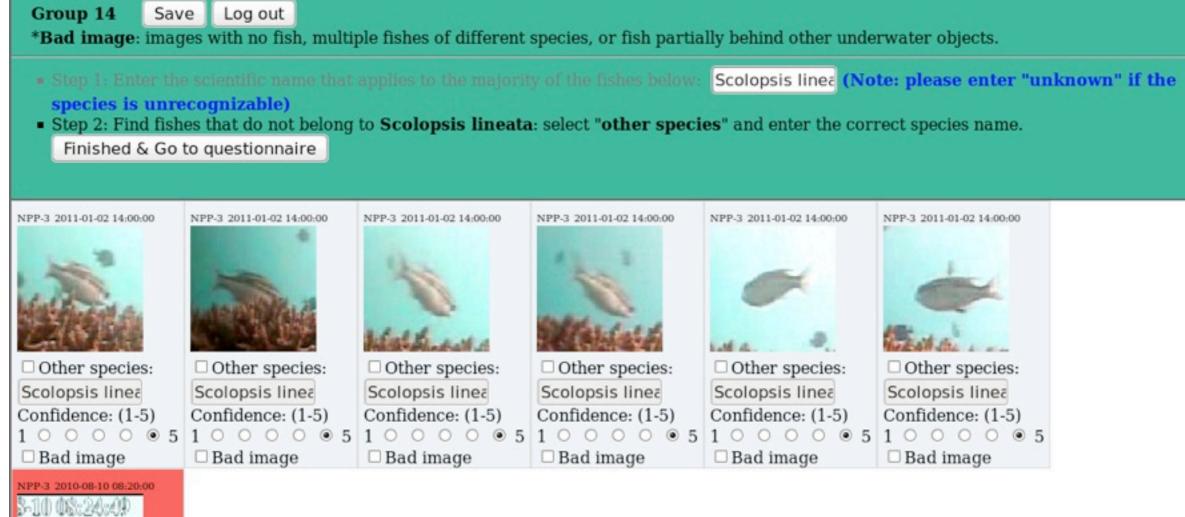


Using non-experts' effort to support expert annotation

- Cluster-based labeling
 - clusters constructed by non-experts
 - instead of labeling every image, experts label the clusters
 - correct the labels of individual images that are wrongly placed in a cluster
 - worst case: labeling every image



Interface (I)





Interface (2)

- A questionnaire after labeling each cluster
 - How difficult was it to recognize the fish (1-5 scale)
 - What makes the recognition difficult?
 - What helps recognition?
 - Are the "other species" related to the main species in the taxonomy tree?
 - What are the most discriminative features of the main species in the cluster?



Expert annotaated data

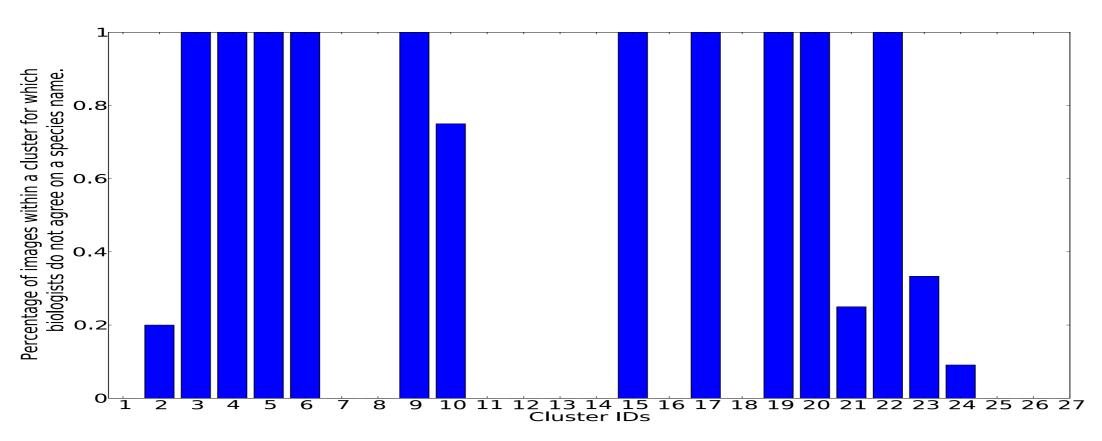
- 3 marine biologists with over 10-25 years' research experience in Taiwan sea area.
- 27 manually constructed clusters
 - For each cluster, at most 30 images are randomly sampled to be shown to the biologists

Agreement among experts - image level

- 82.6% images at least 2 biologists agree on a name
- 56.3% images 3 biologists agree on a name
- Note: sizes of clusters are not evenly distributed

Agreement among experts - per cluster

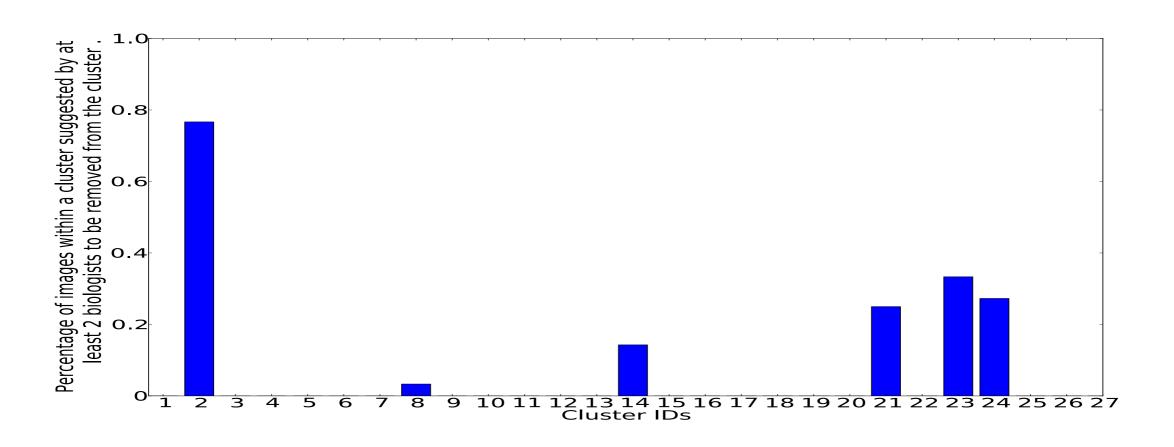
- 9 out of 27 clusters all biologists disagree for every image
- However, for 7 out of the 9 clusters, agreement exists at a family/genus level





Performance of non-expert manual clustering

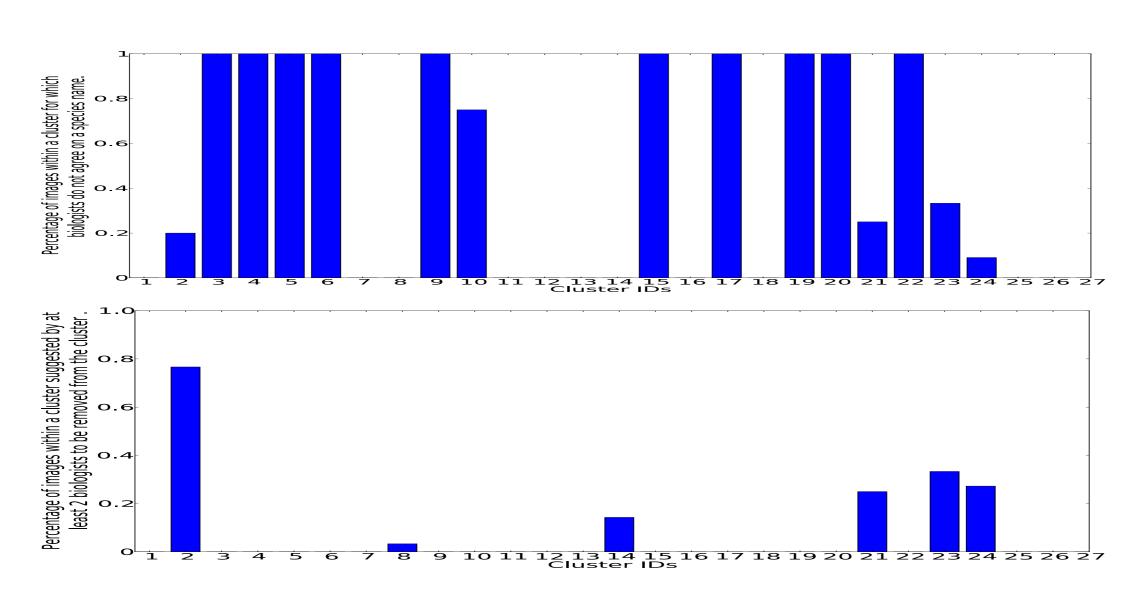
• 6 out of 27 clusters contain "other species"



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Experts vs. non-experts

- Difficult for experts ! = difficult for non-expert
- Clustering != Recognition





Questionnaire

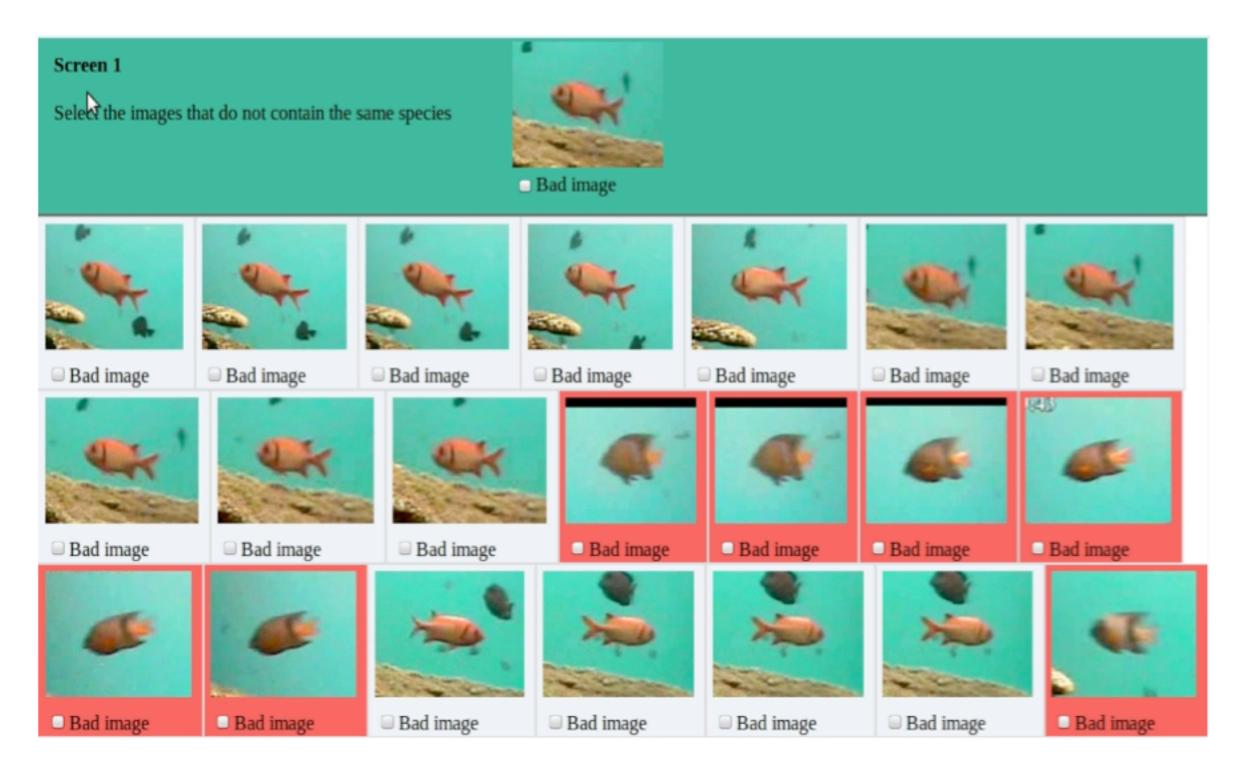
- What makes recognition difficult?
 - 21/27 cases: low resolution
 - 17/27 cases: there exist very similar species
- What helps recognition?
 - 24/27 cases: features of the fish
 - 17/27 cases: experience
 - 5/27 cases: location
 - 3/27 cases: better resolution



Lessons learnt

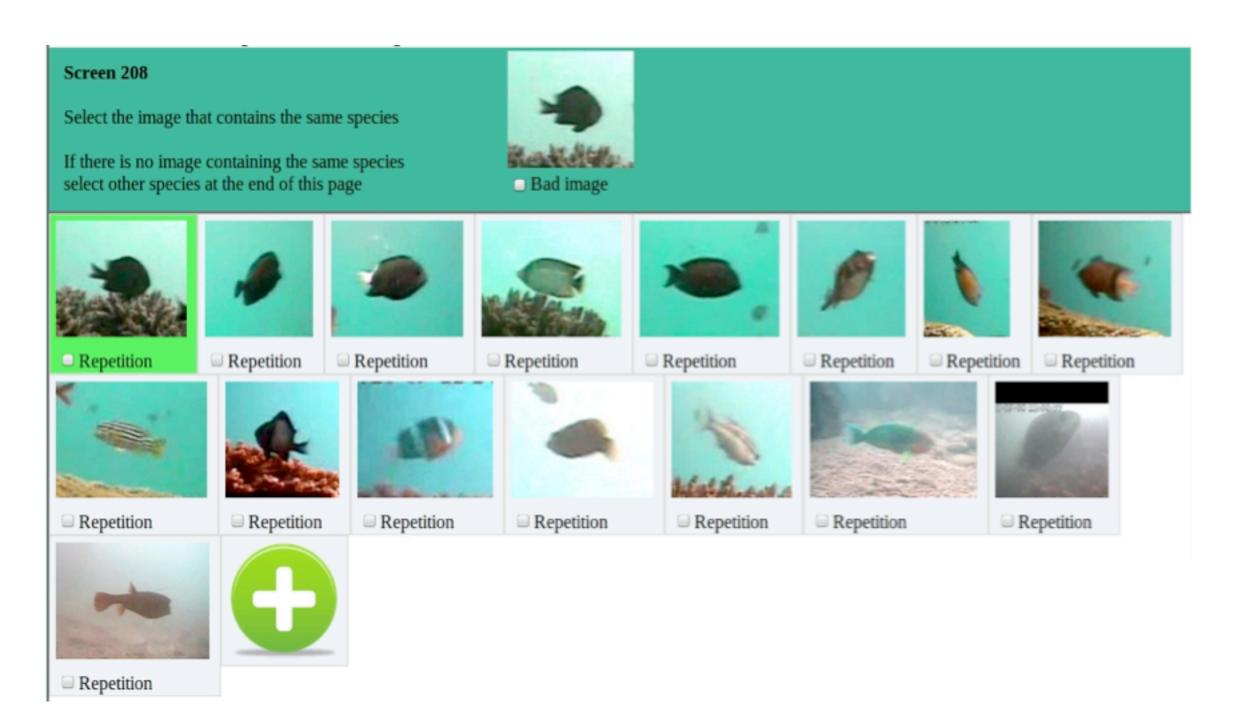
- Cluster-based labeling approach enables experts to label a relatively large amount of images with a limited amount of effort
- Non-experts are able to measure the visual similarity between fish images, thus clustering or classification with visual labels are possible
- Gap exists between clustering and recognition - nontrivial even for experts

Using automatic method to support non-expert annotation: cluster-validation (1)



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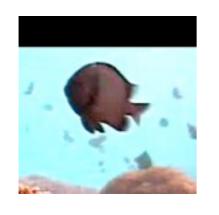
Using automatic method to support non-expert annotation: cluster-validation (2)



CWI



How many different species are there?









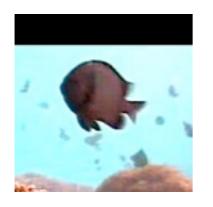








How many different species are there?









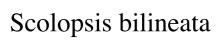
Dascyllus reticulatus

Acanthurus sp.

Zebrasoma scopas

Acanthurus sp.







Scolopsis lineata



Scolopsis lineata



Further steps

- Bridging gap between clustering and recognition
 - Linking clusters to species names
 - Using crowd votes to refine the annotations of images that are visually similar
- Online learning to combine the effort of automatic methods, crowd and experts