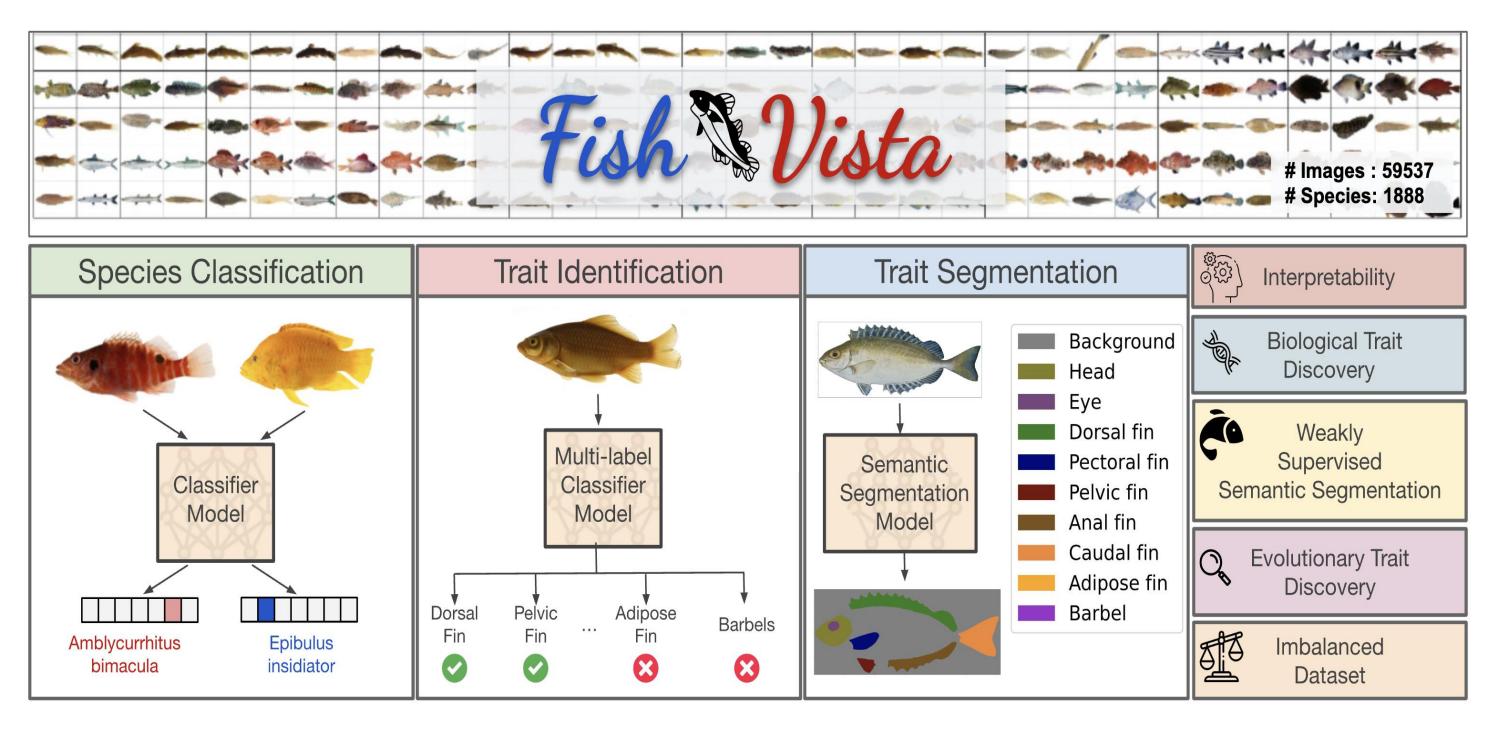
# Fish-Vista: A Multi-Purpose Dataset for Understanding & Identification of Traits from Images

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# Motivation

- The availability of large datasets of organism images combined with advances in AI have significantly enhanced image-based studies of organisms
- However existing organism image datasets have several limitations: • Existing datasets often focus solely on species classification, overlooking tasks related to visual traits of organisms.
- They lack detailed visual trait annotations, such as pixel-level segmentation, crucial for indepth biological research
- They feature organisms in natural habitats, making it difficult to study aquatic species like fish due to poor visual clarity in underwater images
- These limitations hamper the study of aquatic biodiversity and macroevolution



- To address this gap, we introduce the Fish-Visual Trait Analysis (Fish-Vista) dataset: 60K annotated fish images spanning 1900 species
- Supports **biologically relevant** and novel tasks, including species classification, trait identification, and trait segmentation
- Curated through a **sophisticated data processing pipeline** applied to images sourced from various museum collections
- Ensures clear visibility of visual traits, providing **fine-grained labels** for each image • Pixel-level annotations of 9 different traits are available for **4k fish images**, enabling trait segmentation and localization tasks.
- A comprehensive analysis of state-of-the-art deep learning techniques on Fish-Vista is provided to benchmark the dataset.

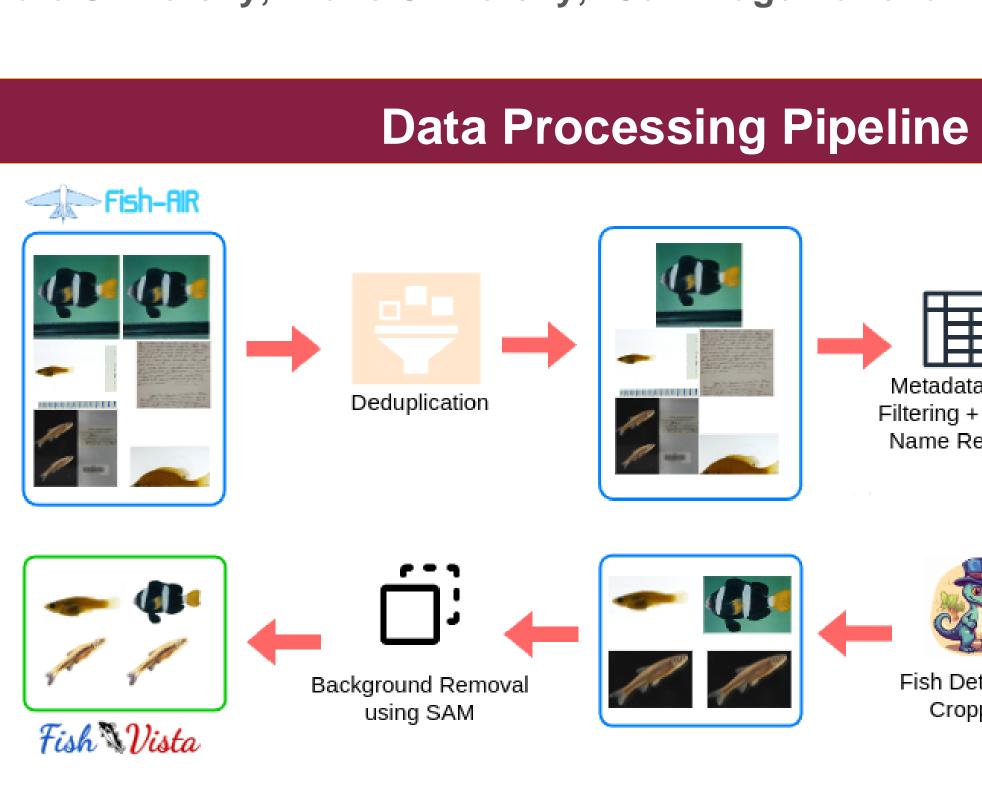
## **Related Datasets**

Related DataSetS							
Dataset	Organism	# Species # Images		Background T	Trait Information Trait Segmentation		
CUB-200-2011 [1]	Birds	200	11,788	Natural Habitat	$\checkmark$	Х	
<b>Birds 525</b> [2]	Birds	525	89,885	Natural Habitat	Х	Х	
NABirds [16]	Birds	555	48,562	Natural Habitat	$\checkmark$	Х	
Stanford dogs [7]	Dogs	120	20,580	Natural Habitat	Х	Х	
Oxford Pet [8]	Cats, Dogs	37	7,349	Natural Habitat	Х	Х	
FathomNet [22]	Marine Species	2244	84,454	Natural Habitat	Χ	Χ	
Ulucan et al. [19]	Fish	9	9,000	Controlled	Х	X	
QUT Fish [4]	Fish	468	3,960	Natural Habitat + Controlled	Х	Х	
DeepFish [9]	Fish	NA	39,766	Natural Habitat	X	Х	
Fish4Knowledge [23]	Fish	23	27,370	Natural Habitat	Х	Х	
FishBase [5]	Fish	35,600	64,000	Natural Habitat	$\checkmark$	Х	
iNaturalist-2021-Fish [18]	Fish	183	46,996	Natural Habitat	X	Χ	
Fish-Vista (Ours)	Fish	1888	59,537	Controlled + Uniform	$\checkmark$	$\checkmark$	









# Manual Filtering of Images

We further manually filter images that do not have all visual traits visible

# **Fish-Vista Tasks and Stats**

### **Fine-grained Species Classification**

Task: Categorizing an image to its species, considering fine-grained differences **Dataset:** 

- Main classification dataset:
- At least 20 images per species; **48k images, 419 species** • Rare species dataset
- 4-20 images per species; **12k images, 1660 species Challenges:** 
  - Extreme long-tailed imbalance
  - Subtle, fine-grained differences between species

### **Trait Identification**

	ADIPUSE FIN
Task: Predicting the presence/absence	
of 4 visible traits that vary across	
species	
Dataset:	PELVIC FIN
<ul> <li>Species-level traits from</li> </ul>	
knowledge bases like	
Phenoscape and FishBase	
<ul> <li>53k images, 682 species</li> </ul>	BARBEL

 51 species are held-out of the training set to evaluate the out-ofspecies generalizability DORSAL F

### Challenges:

- **Trait imbalance:** some traits are rare, like adipose, barbel and multiple dorsal fin
- The trait information is at the species-level, we want to predict on each image
- **Out-of-species generalizability** is more important

### **Trait Segmentation**

**Task:** Involves semantic segmentation of visual traits in fish images

**Dataset:** 

- Manual segmentation annotation of pixel-level traits by expert biologists using the CVAT tool
- Annotation is expensive: 30 images/hour; requires manual inspection
- 4k images, 2.2k species; diverse set of species

### Challenges:

- Trait imbalance: adipose fin and barbel are rarely present
- Some traits are rare and very small, like the barbel





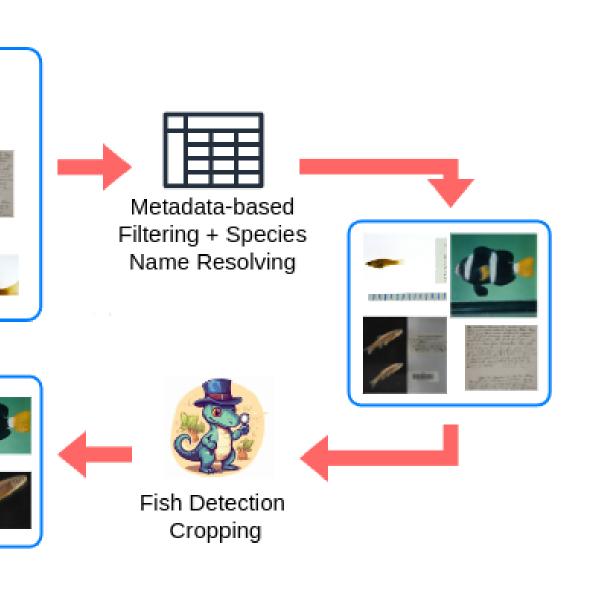


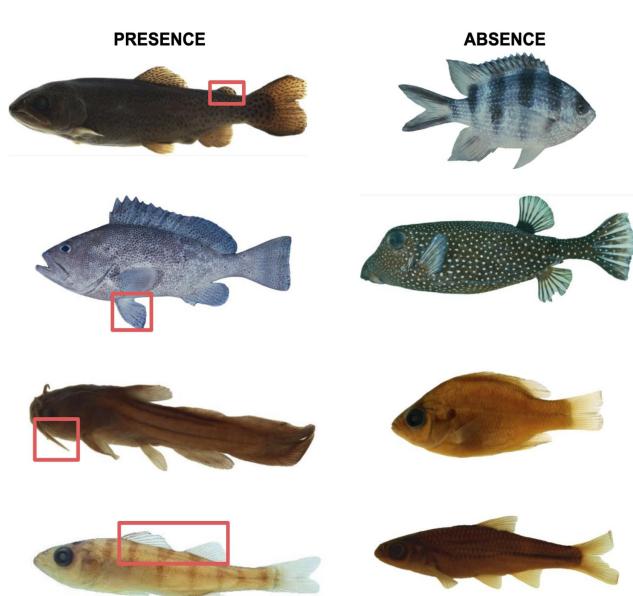






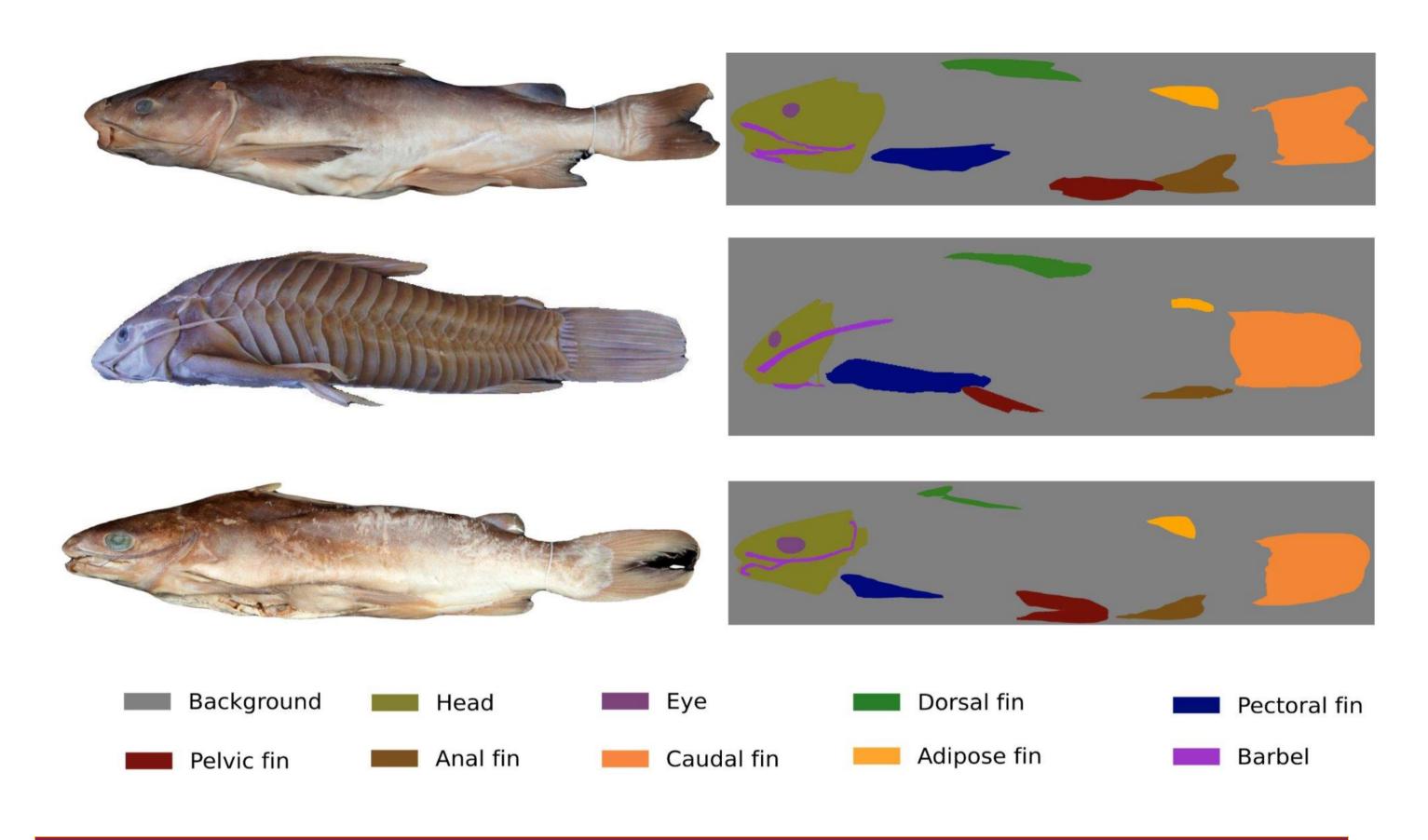






• Some traits are rare and similar looking to other traits, like the adipose fin

### **Examples of Manual Segmentation Annotations**

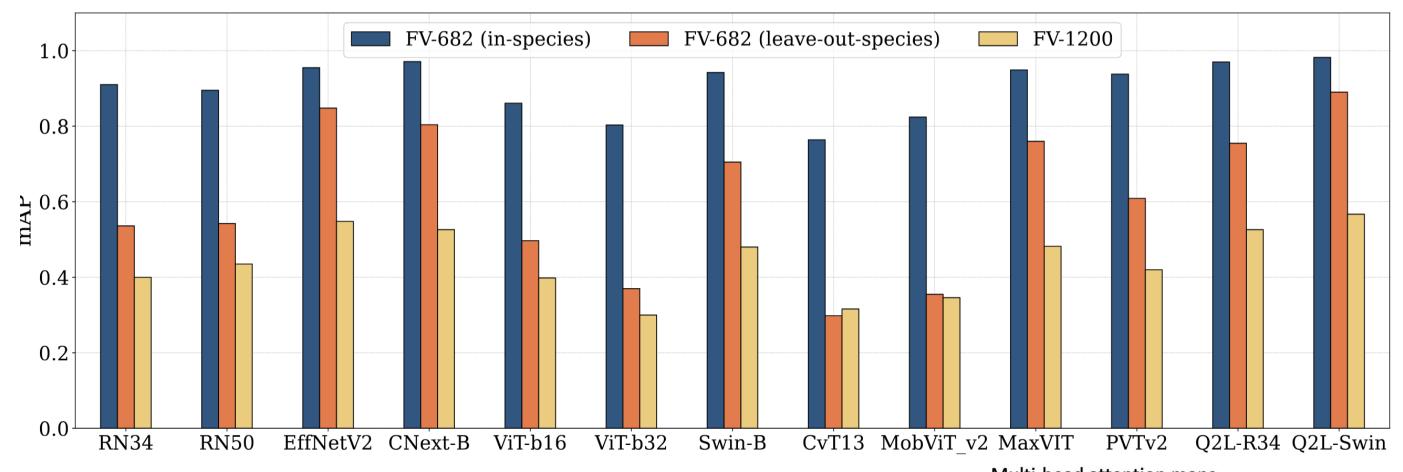


# **Benchmarking Results and Findings**

### **Fine-grained Species Classification**

- performance, while lowering overall performance

### **Trait Identification**



- SOTA methods fail to generalize on out of distribution species
- Models do not 'localize' traits in their predictions

### **Trait Segmentation**

 SOTA methods fail to segment rare and small traits – adipose fin and barbel

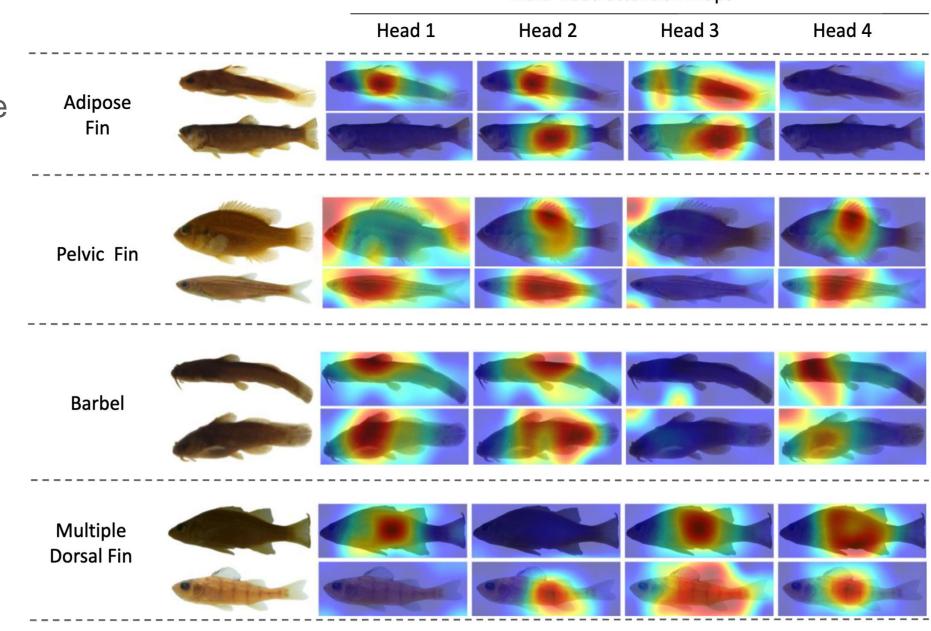
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• SOTA baseline methods achieve poor accuracy on minority (~75%) and rare species (~50%) • Different imbalanced classification methods only slightly improve performance minority

• We evaluate identification baselines on three sets: in-distribution test set, leave-species-out test set, and a manually annotated, diverse leave-species-out set



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