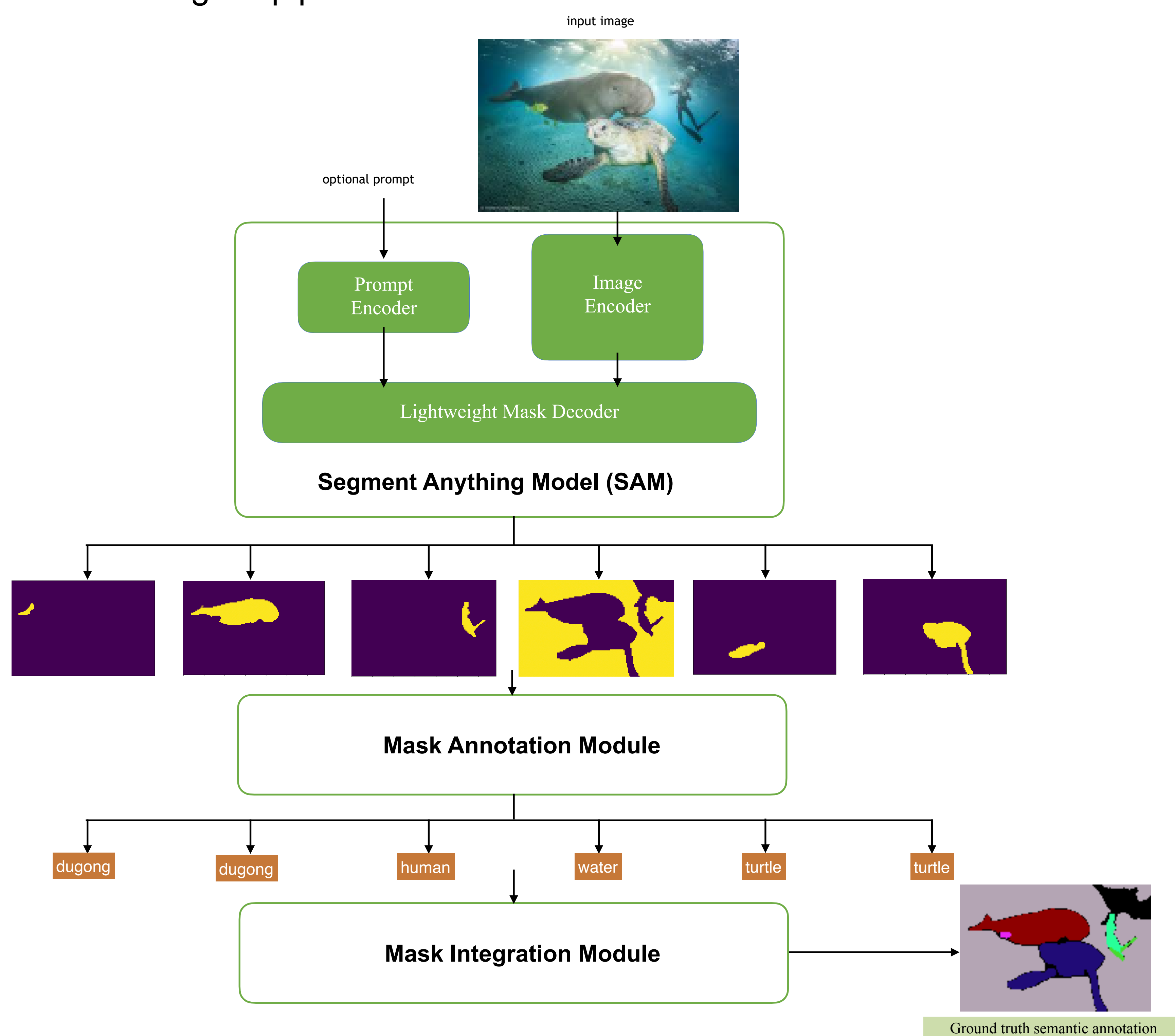


Motivation

- Semantic segmentation is a dense image prediction task where the goal is to label different parts of an image from a set of predefined set of object categories e.g., *table, chair, floor* (indoor) or *sky, car* (outdoor)
- Semantic segmentation has applications in underwater environment exploration, autonomous driving, virtual reality, etc.
- Effective solution for semantic segmentation has two requirements:
 - Deep Neural Network (DNN) models — CNNs or Transformer models
 - Collecting a significant number of images along with dense pixel-wise annotations. These annotated images are used for training the DNNs
- Existing annotation [1] is i) time consuming, ii) laborious, and iii) costly

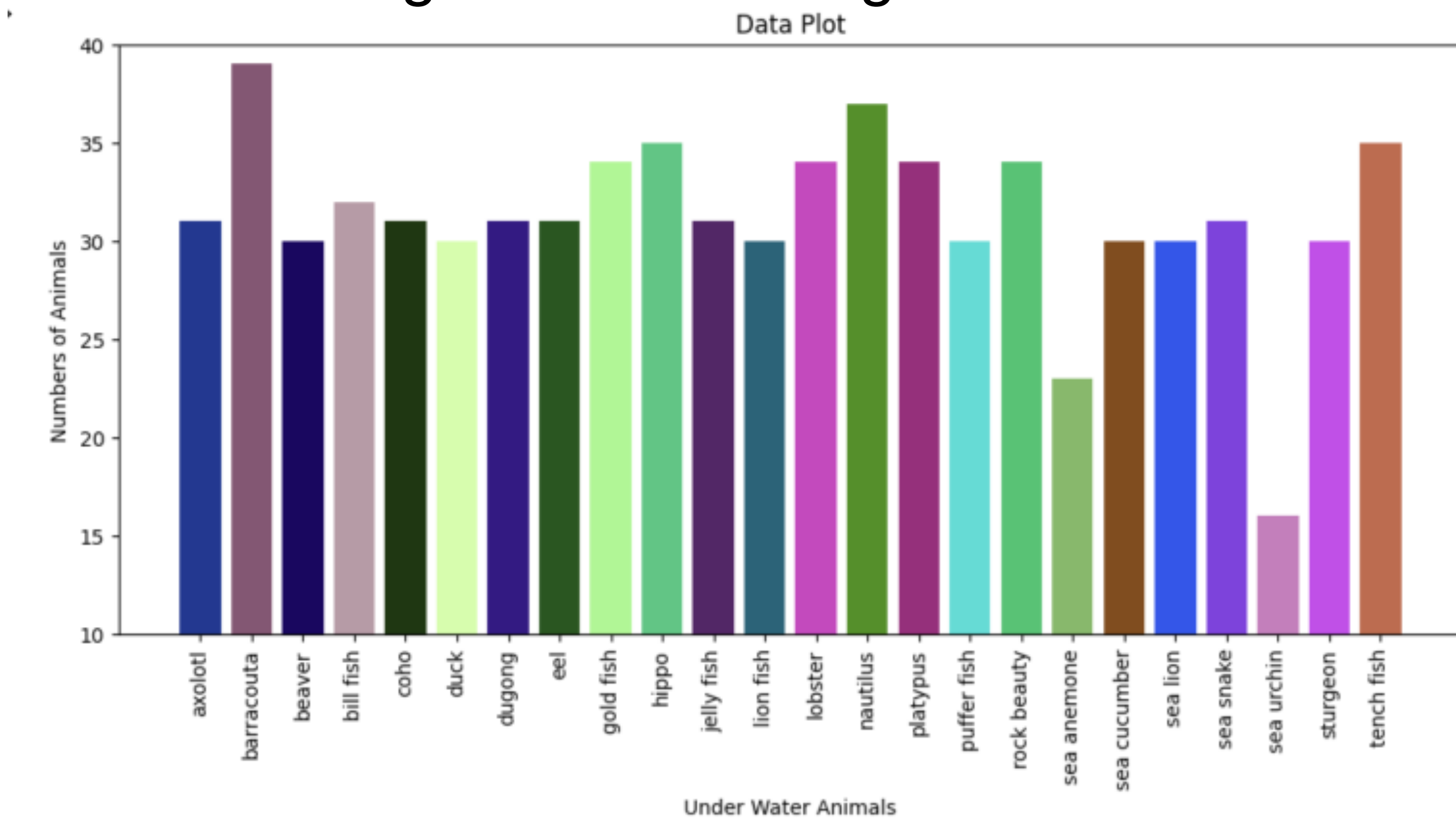
Semi-automatic Annotation Algorithm

- We proposed a semi-automatic algorithm that can annotate every pixel in an image. It only requires entering the name of a segment (vs. laborious manual marking of a region using existing tool [1])
- Our algorithm mitigates some of the limitations of manual annotations — less time-consuming, less labor-intensive, and less costly
- At the core of our algorithm, we have a powerful segment mask generation method called “Segment Anything Model (SAM)” [5]. It can be thought of as a general-purpose segmentation mask generation model — a large language model (LLaMa) equivalent such as ChatGPT. SAM is promotable and has a superior zero-shot performance on a new image distribution
- Our methodological pipeline is shown below:

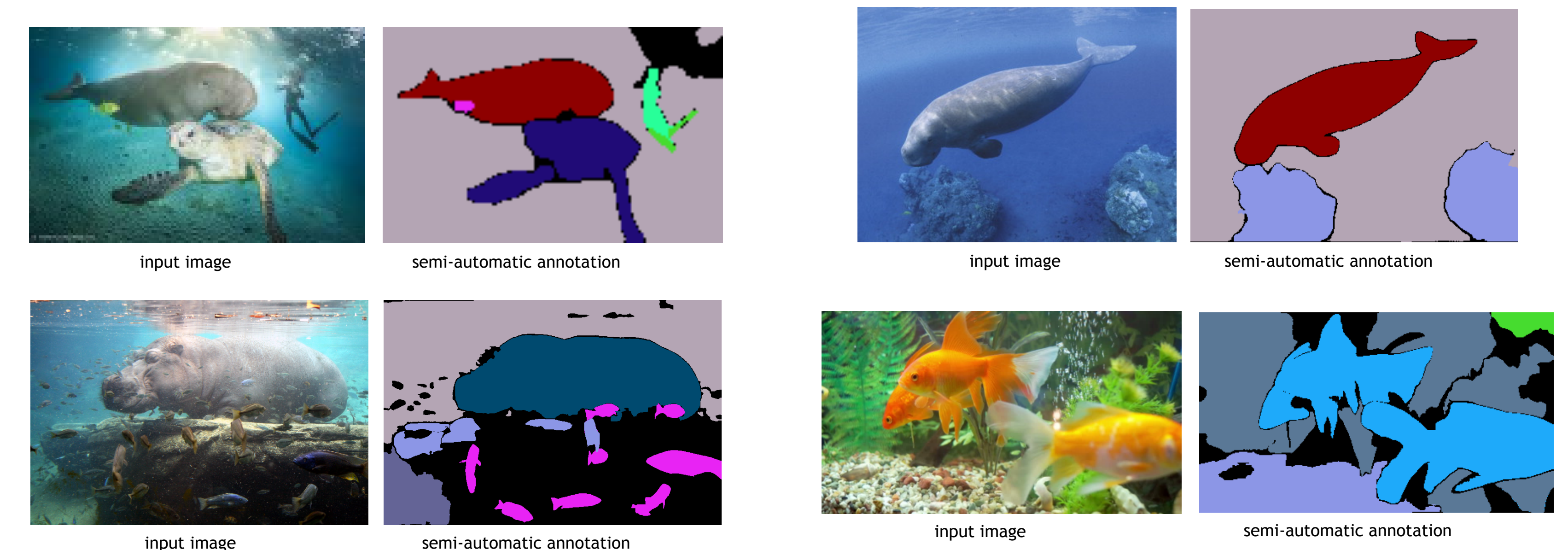


New Dataset for Semantic Segmentation

- We collected images for 24 new animal categories as shown below. It will complement the existing underwater segmentation dataset introduced in [4]



Semi-automatic Annotation Results



Baseline Segment Mask Generation Results

- U²Net [2] can generate segmentation masks but has several limitations:
 - fails to detect all foreground objects
 - merges multiple foregrounds



Future Work

- The semi-automatically annotated images will be used to train DNN semantic segmentation models [3][4]
- Apply our semi-automatic annotation method on other datasets

References

- Label Studio: <https://labelstud.io>
- X. Qin, Z. Zhang, C. Huang, M. Dehghan, O. Zaiane, and M. Jagersand. “U²-Net: Going Deeper with Nested U-Structure for Salient Object Detection”. Pattern Recognition 2020
- M. J. Islam, C. Edge, Y. Xiao, P. Luo, M. Mehtaz, C. Morse, S. Enan, and J. Sattar, “Semantic Segmentation of Underwater Imagery: Dataset and Benchmark,” in IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS), 2020
- I. Kabir, S. Shaurya, V. Maigur, N. Thakurdesai, M. Latnekar, M. Raunak, D. Crandall, and M. Reza, “Few-shot segmentation and Semantic Segmentation for Underwater Imagery” - International Conference on Intelligent Robots and Systems (IROS’23) (under review)
- A. Kirillov, E. Mintun, N. Ravi, H. Mao, C. Rolland, L. Gustafson, T. Xiao, S. Whitehead, A. Berg, W. Lo, P. Doll’ar, and R. Girshick, “Segment Anything” - arXiv’2023