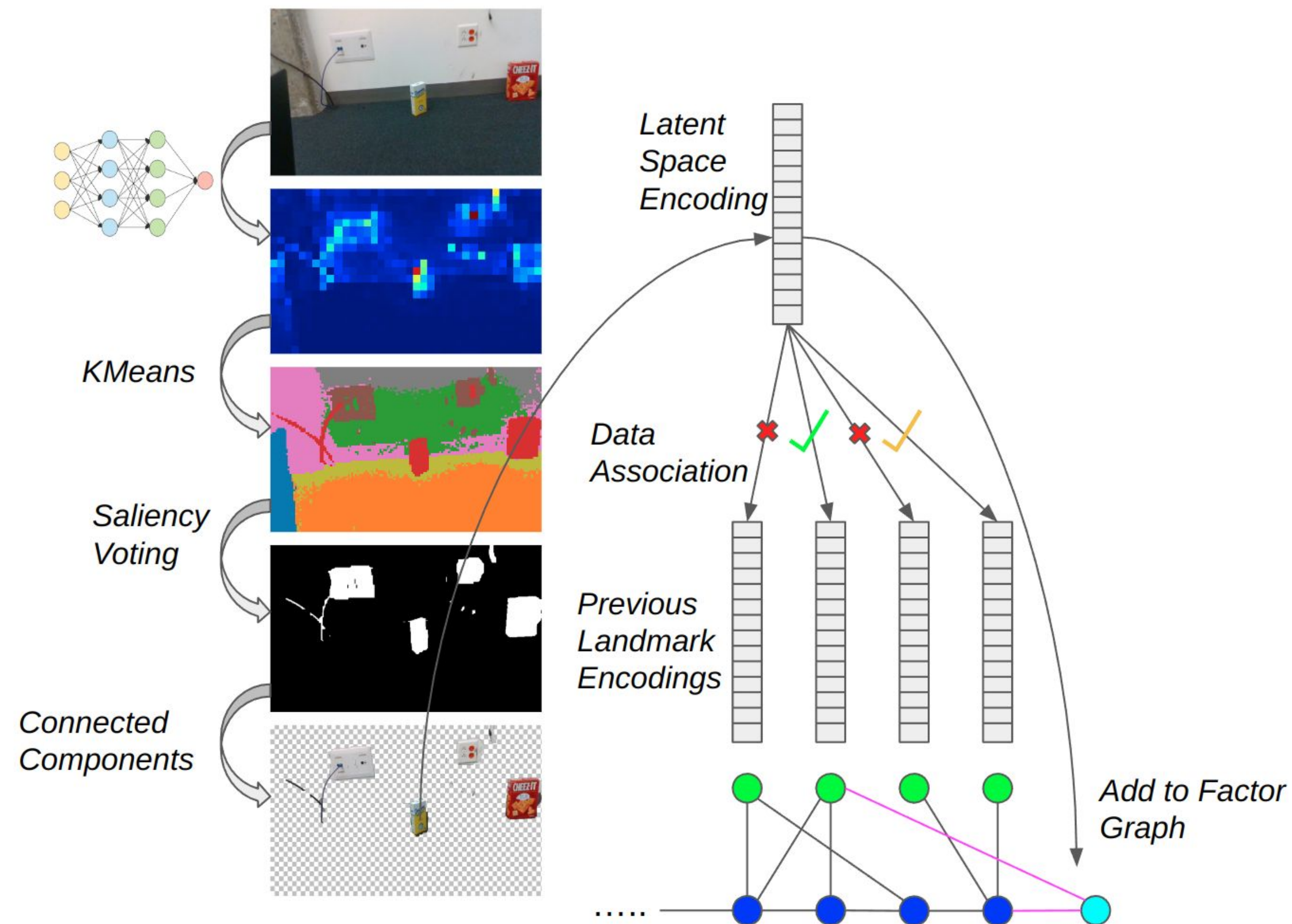


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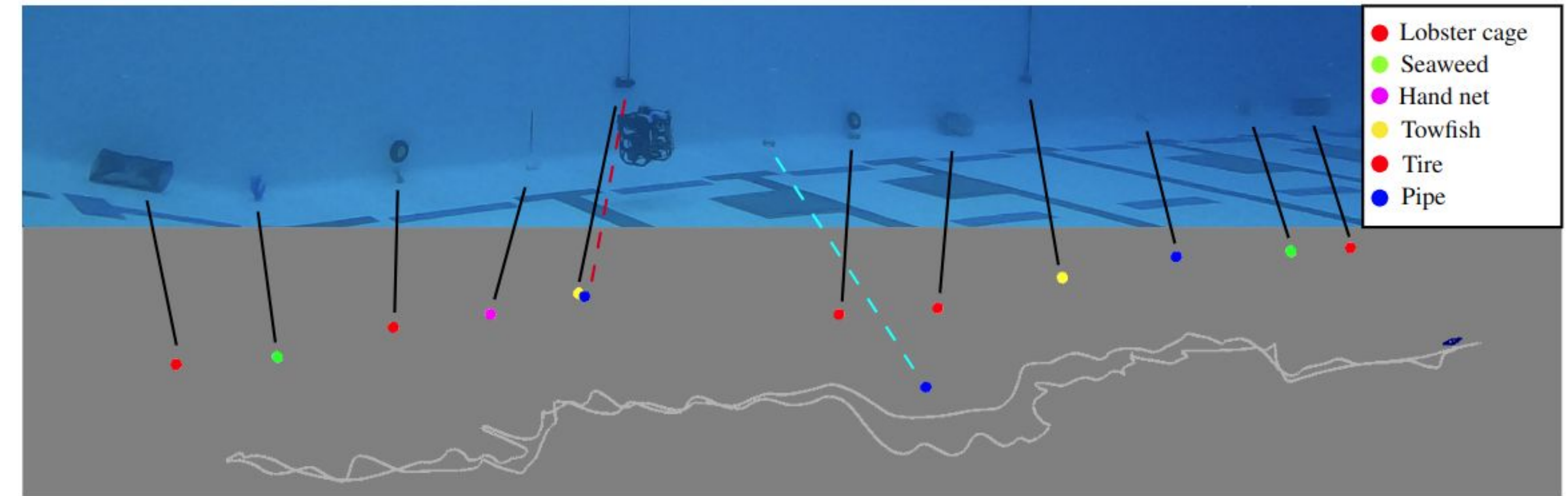
Motivation

- Existing semantic mapping systems are either closed-set (can only detect objects seen in training data) or dense (reasoning at the pixel/feature level rather than object level)
- The goal of this work is to create a method to leverage visual foundation models to build object-level maps while simultaneously enabling robotic systems to robustly localize themselves with such maps

Method



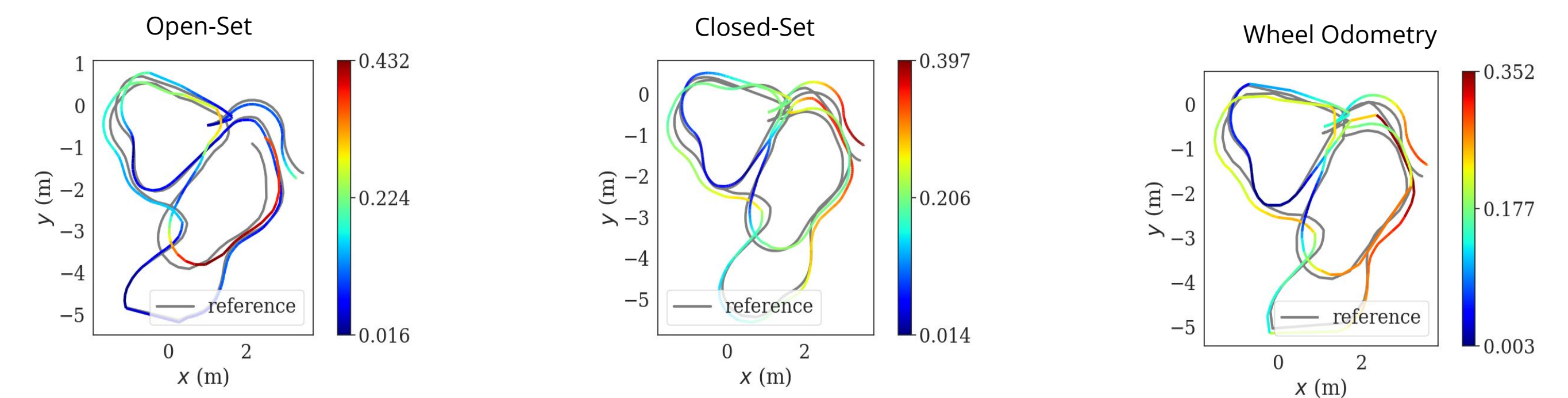
Results



The proposed method has enabled object-level mapping in underwater scenarios, where there is a dearth of labeled training data, and where dense methods fail due to on-board memory requirements. Localization accuracy improves compared to feature-tracking methods, as feature-tracking methods are not robust to underwater lighting effects.

Dataset	Ours (ML)	Ours (EM)	Ours (MM)	Closed Set	Geometric Only (ML)	Wheel odometry only
Pioneer SLAM	0.139	0.184	0.197	0.180	0.180	0.202
Pioneer SLAM 2	0.231	0.342	0.341	0.602	0.472	0.341

Average pose error in meters. The method is evaluated on a public dataset (TUM), demonstrating that our method is able to mitigate odometric drift more effectively than closed-set and geometric methods.



Please visit the project pages below for the full papers, videos, data, and code!



LOSS-SLAM: Lightweight Open-Set Semantic Localization and Mapping



Opti-Acoustic Semantic SLAM with Unknown Objects in Underwater Environments