

ABSTRACT

In uneven illumination scenes, it is difficult to simultaneous localization and mapping (SLAM) because of little information within the dark environment. To solve this problem, first, we propose a novelty visual odometry (VO) which can easily work in dark environments. Then, we propose a framework for multi-robot VO collaboration. Finally, we uses different algorithmic robots to fit different lighting conditions, solving the SLAM task with uneven lighting scenes.

INTRODUCTION

- Background & problem states: Visual SLAM always require the camera to continuously capture bright and clear images to successfully extract feature points or pixel alignment. However, once the acquired images are dark or unclear, the localization accuracy will be significantly decreased.
- Challenges: 1) how to implement SLAM in low illumination environment, and 2) how to build maps in large scenes with minimal resources in a collaborative multi-robot approach?
- **Solutions:** 1) a novelty vision odometry for low illumination environments in uneven illumination scenarios, and 2) a heterogeneous multi-robot collaboration framework with powerful and lightweight robots to avoid resource wastage.

• LIEVO extracts edge features in a dark environment. From the comparison experimental figure, it can be intuitively seen that our method can extract edge features more accuracy and sophistication in the dark environment.

UI-SLAM: A Lightweight Multi-robot Collaborative SLAM System under Uneven Illumination Scenarios

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• Low-level illumination edge-based visual odometry !! Low-lighting-edge (LIEVO): We extract edge features from the images and then perform the localization and mapping by feature matching. The part of edge detection in a dark ||environment uses a deep learning approach.

• Multi-robot collaborative method. In uneven illumination scenarios, the LIEVO algorithm is deployed using a powerful ! NVIDIA Xavier-based drone for low-light environments, while the { LEVO algorithm is deployed using a lightweight Raspberry Pidrone for normal-brightness environments. based The ! collaboration center is a server with higher performance, which \ is mainly used to run global optimization and map fusion.

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	fr1/rpy	0.0800	0.1101	0.0485	0.0511	0.0415	wa
	fr1/teddy	0.0942	0.2325	0.2398	0.3612	0.0753	aco
	fr1/xyz	0.0635	0.0768	0.1005	0.1236	0.0551	OU
	fr2/coke	0.4662	0.4769	0.4689	0.4736	0.4261	aco
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TUM⁴(1, 2)

TUM⁵(1, 2)

CCMSLAM

METHOD

network (LieNet): We use PiDiNet^[1] as a backbone weakly edge method by dark-bright image pair training. tor bright and indicates the detector respectively.

VO RMSE Accuracy: The solute trajectory error (ATE) curacy. It can be seen that r method has the best curacy in the TUM dataset.

ation framework Evaluation: hod is significantly improved ed to the CCMSLAM^[2]. As for the collaboration center, it achieved 45.98% improvement in collaboration accuracy under the uneven illumination scenarios.

CONCLUSIONS

In this paper, the idea of achieving high accuracy SLAM function under uneven lighting scenes was proposed by LIEVO as taken as the positional vision odometry for low illumination environments and heterogeneous multi-robot collaboration framework for more lightweight and efficiency. From the experimental results, it can be seen that the problem of the poor localization accuracy under the uneven illumination has been well solved.

> [1] Su Z, Liu W, Yu Z, et al. Pixel difference networks for efficient edge detection[C]//Proceedings of the IEEE/CVF International Conference on Computer Vision. 2021: 5117-5127. [2] Schmuck P, Chli M. CCM-SLAM: Robust and efficient centralized collaborative monocular simultaneous localization and mapping for robotic teams[J]. Journal of Field Robotics, 2019, 36(4): 763-781.

REFERENCES