

Development of Autonomous Mobile Field Robot – Accuracy Verification of Self-Localization through Simulation -

Takamasa Hayashi

*Graduate School of Computer Science and Systems Engineering, Kyushu Institute of Technology,
680-4 Kawazu, Iizuka-city, Fukuoka, 820-8502, Japan*

**Shintaro Ogawa¹, Yuto Okawachi¹, Tan Chi Jie¹, Janthori Titan¹,
Ayumu Tominaga², Eiji Hayashi¹, Satoko Seino³**

*¹Graduate School of Computer Science and Systems Engineering, Kyushu Institute of Technology,
680-4 Kawazu, Iizuka-city, Fukuoka, 820-8502, Japan*

*²Department of Creative Engineering Robotics and Mechatronics Course, National Institute of Technology,
Kitakyushu College, 5-20-1 Shii, Kokuraminamiku, Kitakyushu, Fukuoka, 802-0985, Japan*

*³Graduate School of Engineering, Kyushu University,
744 Motooka, Nishi-ku, Fukuoka, 819-0395, Japan*

*Email: hayashi.takamasa784@mail.kyutech.jp, tominaga@kct.ac.jp,
haya@ics.kyutech.ac.jp, seino@civil.kyushu-u.ac.jp,
<http://www.kyutech.ac.jp/>*

Abstract

In recent years, the increase in marine debris has become a significant challenge in terms of its collection. Coastal debris, a type of marine debris, can be collected by human hands, but the variety in shapes, and sizes presents limitations to human-only collection efforts. To address this, I focused on developing an autonomous mobile robot, establishing a simulation environment was considered crucial for facilitating smooth progress. This paper focuses on self-localization, an essential aspect for autonomous movement. We replicated an actual coastal cleaning site within the simulation environment and evaluated the accuracy of self-localization using an EKF (Extended Kalman Filter) with multiple sensors.

Keywords: Field Robot, Self-Localization, Gazebo, Simulation, Extended Kalman Filter (EKF)

1. Introduction

In recent years, the increase in marine debris has become a significant challenge in terms of its collection. Coastal debris, a type of marine debris, can be collected by human hands, but the variety in shapes, and sizes presents limitations to human-only collection efforts. To address this, we focused on developing an autonomous mobile robot, establishing a simulation environment was considered crucial for facilitating smooth progress. This paper focuses on self-localization, an essential aspect for autonomous movement. We replicated an actual coastal cleaning site within the simulation environment and evaluated the accuracy of self-localization.

2. Autonomous Mobile Field Robot “BUNKER”

In our previous study, we developed an autonomous mobile field robot [1] based on Kawasaki Heavy Industries’ KFX®90, an all-terrain vehicle (ATV)

powered by a gasoline engine. However, due to the vehicle’s structure, the turning radius was large and stable traveling on steep slopes and rocky terrain was difficult, among other problems. To solve these problems, the platform was changed. The new platform uses Agilix’s BUNKER [2] as shown in Figure 1. The main changes to the platform are that the traveling mechanism has two opposing wheels and the drive wheels are crawlers. This enables the platform to make super-clever turns, improving maneuverability composed to conventional platforms. In addition, the robot can climb hills with a slope angle of 36°, which is expected to enable it to run at higher speeds. The robot is equipped with an RGB-D sensor and 3D LiDAR as a visual system for autonomy. An encoder is mounted inside the vehicle body to measure the rotation speed of the wheels.



a. KFX@90 b. BUNKER
Fig. 1. Platform Overview

2.1. Simulation Construction

In this study, we not only represented the Agilex BUNKER in the simulation environment but also recreated the coast of Hokuto Mizukumi Park in Munakata City, Fukuoka Prefecture, where actual beach cleaning is performed. We used 3D mapping technology developed by prior research using drones. For the simulation environment, we employed Gazebo, a 3D dynamic simulator capable of efficiently and accurately simulating groups of robots in complex indoor and outdoor environments. Gazebo can also be integrated with ROS (Robot Operating System) [3]. The projected BUNKER and the coast on Gazebo are shown in Figure 2.



a. BUNKER b. coast
Fig. 2. Built simulation environment

2.2. EKF-Based Self-Localization System

The posture measurement of mobile robots is achieved through various sensors and methods. However, the measurements obtained from these sensors are not true values but are considered to contain errors. In this study, we implemented the integration of odometry using the Extended Kalman Filter (EKF) shown in Figure 3, to achieve robust and low-error self-localization. We used wheel odometry, RGB-D odometry, and LiDAR odometry. The system was designed to accept any of these, individually or in combination, and produce an integrated odometry output.

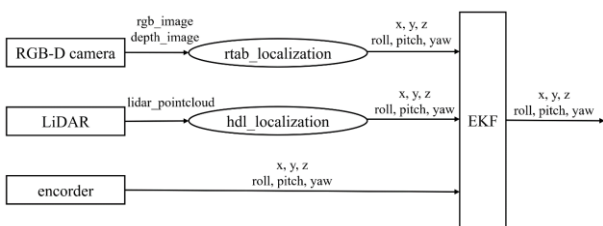


Fig. 3. Input/output data to/from EKF

3. Experiment

In the coastal environments shown in Figure 4 and Figure 5, the initial position of the robot was set as the origin of spatial coordinates, and from there, three destinations were set at (40, -5), (40, 0), (40, 5). The robot was remotely operated to each destination five times to extract the necessary odometry information. For self-localization, both individual odometries and multiple odometries integrated via EKF were implemented, resulting in seven estimation patterns. Mean Absolute Error (MAE) was used for error calculation.

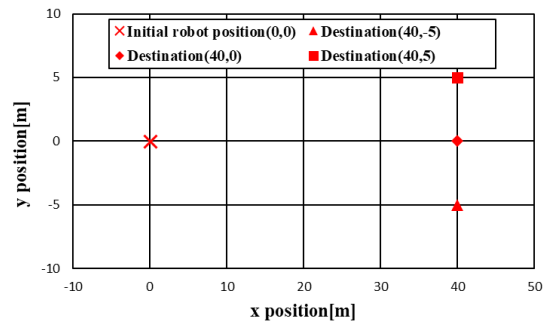
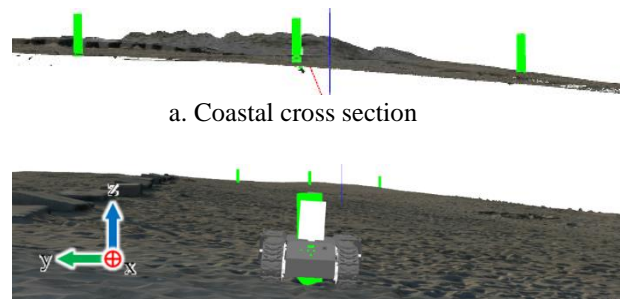


Fig. 4. Test environment



a. Coastal cross section
b. View from initial position
Fig. 5. Experiment in a simulation environment

3.1. Result

Table 1 summarizes the errors in self-localization results in the x-axis direction. In Pattern 4, the error was minimal, with the odometry integrating wheel odometry and RGB-D odometry via EKF achieving the highest accuracy in self-localization. Additionally, focusing on Pattern 2, the RGB-D odometry alone also showed generally similar values. From Pattern 3, it was observed that the LiDAR had an error of about 20 m, indicating it was not very useful in the simulated environment replicating an actual beach.

Table 1. Error in Self-Localization along the X-axis [m]

	Wheel	RGB-D	LiDAR	(40,5)	(40,0)	(40,-5)
1	○			11.06	1.22	0.35
2		○		0.08	1.06	1.44
3			○	21.66	25.25	19.50
4	○	○		0.85	0.47	0.86
5	○		○	13.73	5.42	4.47
6		○	○	10.36	10.07	8.27
7	○	○	○	3.20	4.33	5.54

Table 2 summarizes the errors in self-localization results in the y-axis direction. Compared to the results in the x-axis direction, there were significantly larger discrepancies overall. In every pattern, high-accuracy self-localization was not achieved. While some input destinations yielded accurate results, a stable trend was not observed. Among them, the odometry that integrated wheel odometry and RGB-D odometry via EKF had the smallest error.

Table 2. Error in Self-Localization along the X-axis [m]

	Wheel	RGB-D	LiDAR	(40,5)	(40,0)	(40,-5)
1	○			17.90	3.71	1.17
2		○		2.35	12.93	6.09
3			○	15.11	19.68	13.11
4	○	○		2.49	12.68	5.44
5	○		○	14.60	3.14	4.13
6		○	○	7.30	13.64	3.94
7	○	○	○	2.96	12.50	5.28

3.2. Consideration

Throughout this experiment, RGB-D odometry was found to have an error and greatly contributed to the improvement of self-localization accuracy. However, the accuracy of LiDAR odometry was considerably low in coastal environments like those in this study, lacking distinct landmarks for reference. The experimental environment featured a slope descending to the right from the travel direction, as depicted in Figure 5, leading to vehicular lateral skidding, which mainly affected y-axis self-localization. The introduction of an Inertial Measurement Unit (IMU) is proposed as a solution for more accurate vehicle posture determination, with prior research suggesting its substantial utility in enhancing results.

4. Conclusion

In this study, we developed a simulation environment for an autonomous mobile field robot and conducted accuracy verification of self-localization. As a result, odometry that integrated wheel odometry and RGB-D odometry using RGB-D and EKF was found to achieve more accurate self-localization. In the future, we aim to improve the accuracy of self-localization by adding an

IMU. we will also conduct verification on sandy beaches different from those in this study.

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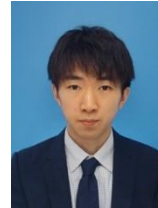
Authors Introduction

Mr. Takamasa Hayashi



He received his Bachelor's degree in Engineering in 2023 from intelligent and Control Systems, Kyushu Institute of technology in Japan. He is currently a Master student in Kyushu Institute of Technology and conducts research at Hayashi Laboratory.

Mr. Shintaro Ogawa



He received his Bachelor's degree in Engineering in 2022 from intelligent and Control Systems, Kyushu Institute of technology in Japan. He is currently a Master student in Kyushu Institute of Technology and conducts research at Hayashi Laboratory.

Mr. Yuto Okawachi



He received his Bachelor's degree in Engineering in 2023 from intelligent and Control Systems, Kyushu Institute of technology in Japan. He is currently a Master student in Kyushu Institute of Technology and conducts research at Hayashi Laboratory.

Mr. Tan Chi Jie



He received his Master of Creative Informatics from the Department of Computer Science and Systems Engineering, Kyushu Institute of Technology, Japan in 2023. He is currently a Doctoral student at Kyushu Institute of Technology and conducts research at Hayashi Laboratory.

Mr. Janthori Titan



He received his Bachelor of Engineering in Robotic Engineering and Automation System from the Department of Production Engineering, Faculty of Engineering, King Mongkut's University of Technology North Bangkok in 2022. He is currently a Master student at Kyushu Institute of Technology and conducts research at Hayashi Laboratory.

Assist. Prof. Ayumu Tominaga



He is a professor in Department of Creative Engineering Robotics and Mechatronics Course at National Institute of Technology Kitakyushu College. He received the Ph.D. (Dr. Eng.) degree from Kyushu Institute of Technology in 2021. His research interests include Intelligent mechanics, Mechanical systems and Perceptual information processing.

Prof. Eiji Hayashi



He is a professor in the Department of Intelligent and Control Systems at Kyushu Institute of Technology. He received the Ph.D. (Dr. Eng.) degree from Waseda University in 1996. His research interests include Intelligent mechanics, Mechanical systems and Perceptual information processing.

He is a member of The Institute of Electrical and Electronics Engineers (IEEE) and The Japan Society of Mechanical Engineers (JSME).

Associate. Prof. Satoko Seino



She is an associate professor in Graduate School of Engineering, Kyushu university. She received the Ph.D. (Eng.) from Kyushu University. Her research interests include coastal and riverine environmental conservation.