# **Spatial Mapping**

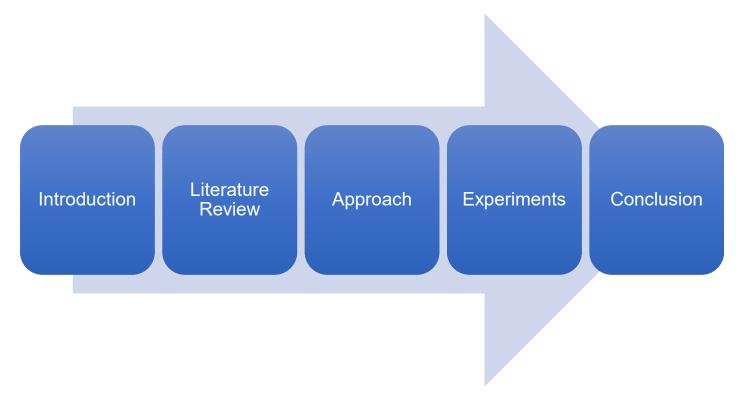
Research Project

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## Content



### Introduction

 Maps have been used for centuries by humans for centuries trying to navigate around the earth.

• First evidence – 1150 BC, in Turin.

- Why do we need maps in robotics?
  - Localization
  - Path Planning

• ..



### In this work...

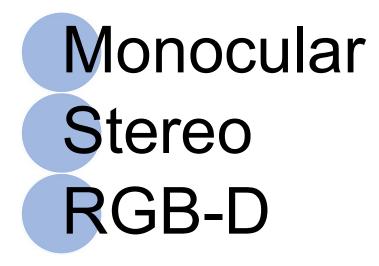
- The core idea of this work is to investigate a few suitable Visual SLAM solutions for map building.
- Choice of Visual SLAM Ease of setup and better (comprehensible) map representation.

### **Fundamentals**



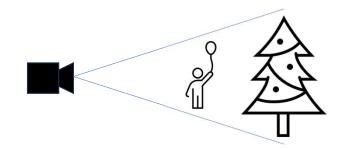
### Sensors

Visual SLAM – Vision sensors (or Cameras)

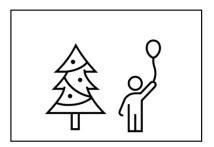


### Monocular Sensors

 Monocular sensors are cameras that images the scene.



- Advantage They are easily accessible and cost effective
- Disadvantage They lack scale
  - No accurate way of determining ranges.



# Some popular examples

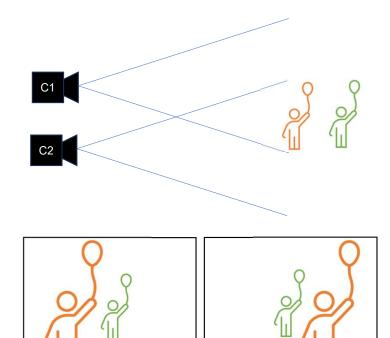




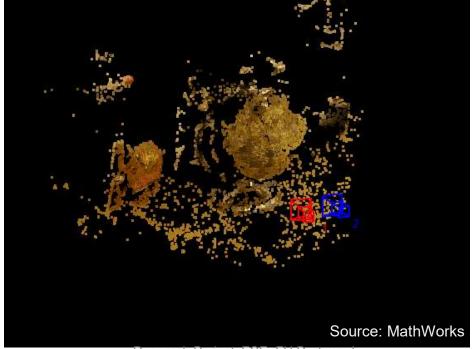
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### Stereo Sensors

- It is a pair of monocular sensors, separated by a distance (baseline).
- Advantage Preserve Scale
  - Depth can be calculated mathematically.
- Disadvantage Requires computation
  - But often off-the-shelf cameras provide depth information.



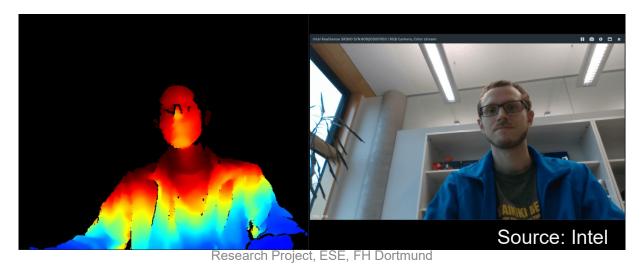




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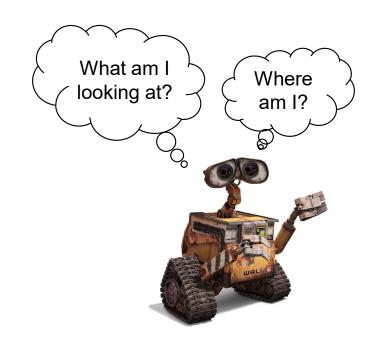
### **RGB-D Sensors**

- As the name suggests it provides both color and depth information.
- Depth is hardware estimated
  - Structured Light and ToF measurements provide very high accuracy.



### The SLAM Problem

- As contained in the name, SLAM is Simultaneous Localization and Mapping.
- The SLAM problem is to construct a map of the environment and localize the robot in this environment, simultaneously.



### The SLAM Problem

Mathematically, it is formulated as

$$p(x_t, m \mid Z_t, U_t)^1$$

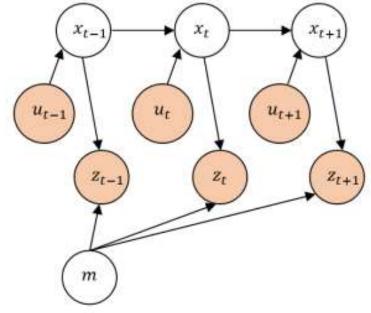
#### Where,

 $oldsymbol{\mathcal{X}}_t$  is the robot position at any time t

m is the map

 $Z_t$  is the sequence of measurements at time t

 $U_t$  is the odometry at time t



Source: [SK16]

# **SLAM Paradigms**

#### EKF SLAM

 The robot state and the measurement states are represented as probabilistic distributions and the unknowns are estimated using recursive filtering.

#### Particle Filter SLAM

 The robot state (posterior) represented through a set of particles, representative of the best guess of the true robot state. By collecting many such approximations into a set of good guesses, the algorithm estimates the robot state.

#### Graph-based SLAM

 The robot state and the map is represented as a graph interconnected by the means of observation. The two kinds of nodes in the graph represent the robot pose or a landmark on the map. The edges of the graph represent an observation.

### **Evaluation Metrics**

#### Relative Pose Error

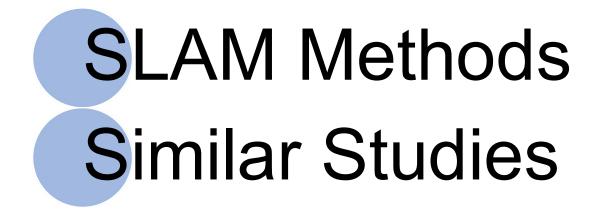
- The relative pose error, as the name suggests, is a measure of the local accuracy of the poses over a fixed time interval.
- Therefore, it is a good indicator of the drift of the trajectory which is one
  of the major challenges any visual odometry system faces.

#### Absolute Trajectory Error

- This method compares the absolute distances between the estimated and the ground truth trajectory.
- It provides a global consistency of the estimated poses.

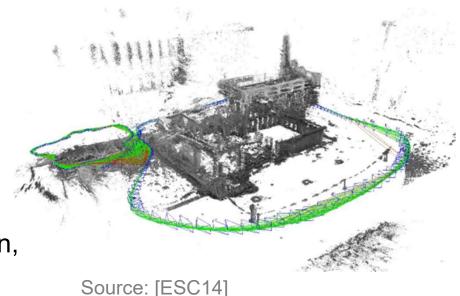
### Literature

The literature survey is divided into two parts in the following manner.



### LSD SLAM

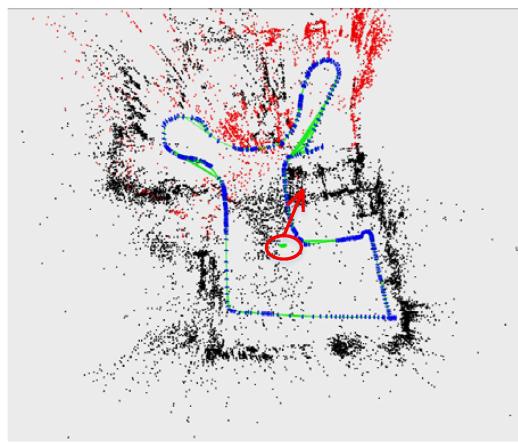
- Large-Scale Direct SLAM
- LSD SLAM1, first introduced in [ESC14] is a monocular VSLAM algorithm developed by Engel, Schöps, and Cremers.
- This is a direct SLAM algorithm
  - Advantages: better accuracy, robustness to changes in illumination, low-texture environments.



### ORB SLAM 2

- ORB-SLAM2 introduced in [MT17] is a real-time VSLAM algorithm developed by Mur-Artal and Tardos at the University of Zaragoza in Spain.
  - It is an improved version of ORB-SLAM (2015)
- The major improvement over its predecessor that this method offers is the support for stereo and RGBD inputs.
- It uses a bag-of-words (BoW) approach to represent the visual vocabulary of the environment.

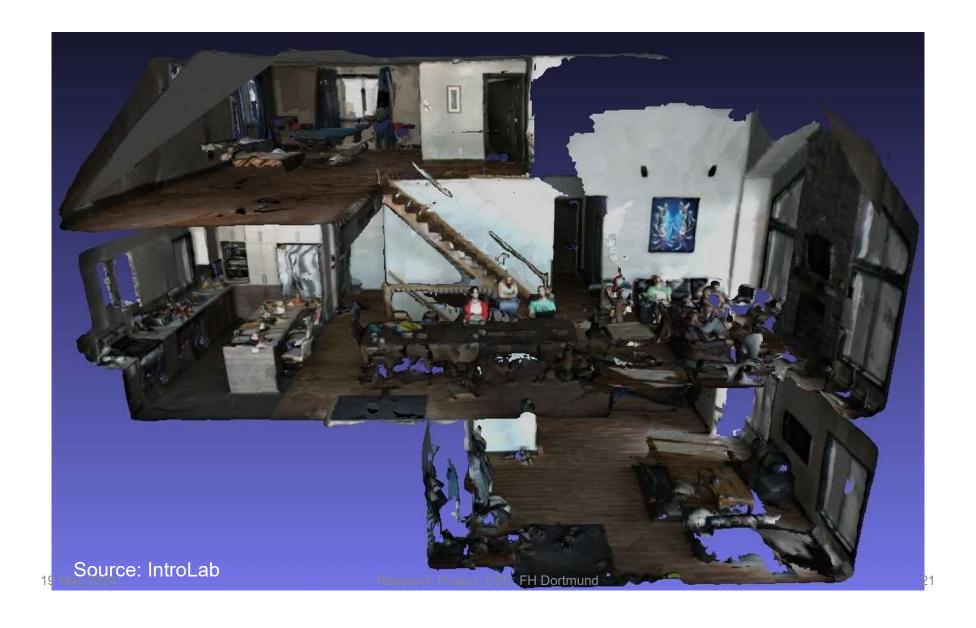




Source: MDPI, Rapid Relocation Method for Mobile Robot Based on Improved ORB-SLAM2 Algorithm

# RTAB Map

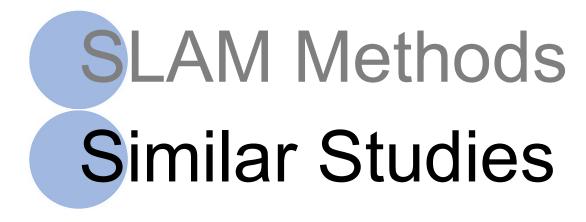
- Real Time Appearance Based Mapping [LM13] introduced by Labbé and Michaud in 2013.
- Originally intended for appearance-based loop closure detection approach with long-term memory management.
- Further extended by the authors in [LM19] as an open-source solution for Visual and LiDAR SLAM.
- Advantages
  - Better loop-closure and long-term localization.
  - advanced memory management scheme which facilitates fast and efficient relocalization.





### Literature

The literature survey is divided into two parts in the following manner.



# Similar Studies

| Work  | Source   | YEAR | Compared Algo.  |  |
|---|----------|------|---|--|
| A Comprehensive Survey of<br>VSLAM Algorithms   | [Mac+22] | 2022 | KinectFusion, SLAM++,<br>RGBDSLAMv2, DVO, ORB<br>SLAM 2 |  |
| An evaluation of ROS-<br>compatible stereo VSLAM<br>methods on a NVIDIA Jetson<br>TX2 | [Giu+19] | 2019 | ORB SLAM 2, SPTAM, RTAB-Map                             |  |
| Mapping and Navigation for Indoor Robots under ROS: An Experimental Analysis          | [DXG19]  | 2019 | GMapping, RTAB-Map                                      |  |
| Comparison of Various SLAM<br>Systems for Mobile Robot in an<br>Indoor Environment    | [FA18]   | 2018 | LSD SLAM, ORB SLAM 2,<br>RTAB-Map                       |  |

#### A Comprehensive Survey of VSLAM Algorithms [Mac+22]

- The work by Macario Barros et al. pits the ORB SLAM 2 as a candidate amongst DVO, RGBDSLAM, KinectFusion, and other prominent methods.
- And according to this study, ORB SLAM 2 offers good performance in terms of localization performance, loop-closing etc. But it falls short in case of map quality (density) which is a requirement

# An evaluation of ROS compatible stereo VSLAM methods on a NVIDIA Jetson TX2 [Giu+19]

- The work by Giubilato et al. compares the performance of ORB SLAM 2 and RTAB-Map on a embedded platform and therefore provides a good outlook on how the methods would be implemented on a robotic system.
- It also discussed the possible improvements that can be made to the SLAM pipeline, in terms of preprocessing input and parallelising the core processes of the SLAM system.

# Mapping and Navigation for Indoor Robots under ROS: An Experimental Analysis [DXG19]

- The paper by Da Silva, Xavier, and Gonçalves compares RTAB-Map and Gmapping in simulation and studies the configurability of both systems and how it can reflect in the overall mapping output of the system.
- This paper also discusses map reuse, which is a point of interest to the research introduced in this thesis

# Comparison of Various SLAM Systems for Mobile Robot in an Indoor Environment [FA18]

 The paper by Filipenko and Afanasyev compares all 3 proposed methods in this thesis – LSD SLAM, ORB SLAM 2, and RTAB-Map. It studies the indoor localization performance and pits RTAB-Map as a good candidate for the use-case.

# Approach

# Evaluate localization and mapping accuracy.

- Dataset based evaluation
- Handheld Mapping of an object and area



# Setup a system on ROS

Simulation Setup

## Requirements

- The primary requirement of this system is to produce a 3D map of its environment or the defined work area that can be visualized by a human operator.
- The 3D map is estimated in good quality, with resolution in the range of 1cm voxel grids with an error limit of around 10-15%.
- 2D maps are also to be created, which can then be used by other robots to navigate around the work area

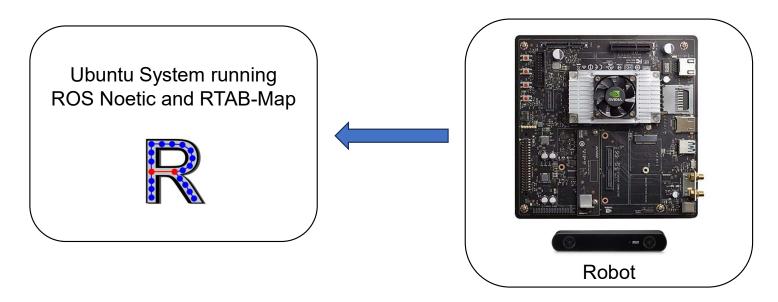
# Experiments on the Dataset

- TUM RGBD dataset is used to benchmark the system.
- Estimated trajectory is compared with the GT so that the accuracy of the localization system can be evaluated.

|                               | RTAB-M   | Iap                  | ORB SLAM 2 |                      |  |
|-------------------------------|----------|----------------------|------------|----------------------|--|
|                               | RMSE (m) | $\sigma(\mathrm{m})$ | RMSE (m)   | $\sigma(\mathrm{m})$ |  |
| fr2/desk                      | 0.029    | 0.00                 | 0.065      | 0.05                 |  |
| $fr3/long\_office\_household$ | 0.018    | 0.01                 | 0.070      | 0.05                 |  |
| $fr2/pioneer\_360$            | 0.038    | 0.00                 | 1.059      | 0.53                 |  |
| $fr2/pioneer\_slam2$          | 0.045    | 0.02                 | 1.709      | 0.18                 |  |

# **Experimental Setup for Mapping**

- The experiment is set up on a computer and an Nvidia Jetson.
- The camera used is a StereoLabs ZED2



# Experiments with ZED2

The experiments with the camera is again divided into two:

Mapping of an object.

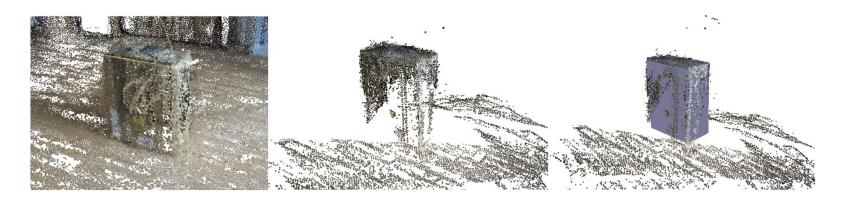
Mapping of an area.

# Mapping an Object

 A scan of an object (box) is performed using the pipeline and a point cloud of the resultant scene is exported from the RTAB-Map application.

- The point cloud is cleaned up in MeshLab to facilitate measurement.
- Measurement is done using the inspect functionality in MeshLab.
- The point cloud is also visually compared against a synthetic 3D mesh of the original box.

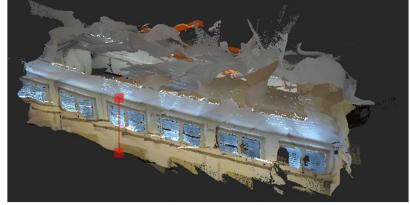
# Results

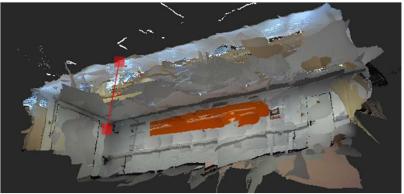


|      | Length (m) |          | Breadth (m) |          | Height (m) |          | Error |
|------|------------|----------|-------------|----------|------------|----------|-------|
|      | Measured   | Original | Measured    | Original | Measured   | Original | %     |
| Box1 | 0.35       | 0.34     | 0.20        | 0.19     | 0.45       | 0.46     | 2.84  |
| Box2 | 0.33       | 0.27     | 0.22        | 0.23     | 0.18       | 0.16     | 11.27 |

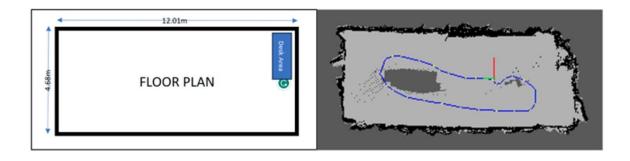
# Mapping an Area

- Next the pipeline is used to map an indoor area.
- The GT here is the measurement of the room lengths performed using a Bosch PLR50.

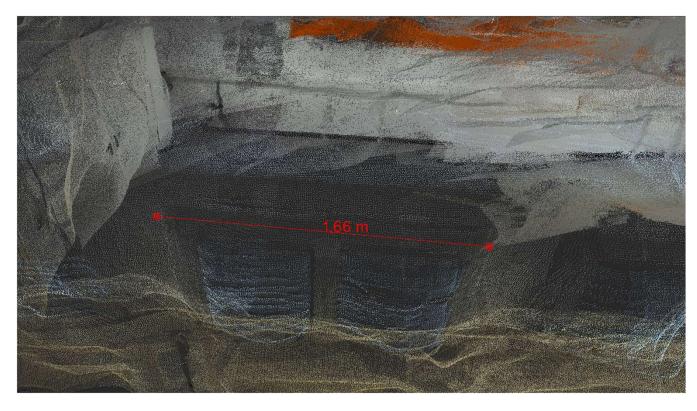




# Results



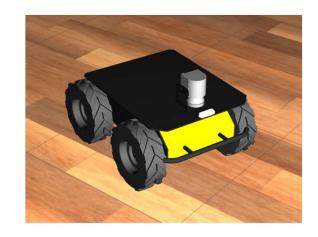
|                             | Length (m) |          | Breadth (m) |          | Height (m) |          | Error |
|-----------------------------|------------|----------|-------------|----------|------------|----------|-------|
|                             | Measured   | Original | Measured    | Original | Measured   | Original | %     |
| Room1                       | 12.11      | 12.01    | 4.31        | 4.68     | 2.35       | 2.45     | 4.55  |
| $\operatorname{Corridoor1}$ | 18.43      | 19.74    | 1.11        | 1.58     | 2.3        | 2.45     | 6.64  |





# Setting up SLAM in Simulation

- As a preliminary to building the actual robotic system, the system is prototyped and demonstrated on Gazebo.
- The Husky is chosen as a platform for the prototype implementation of the system
- The depth camera plugin is developed using the OpenNI plugin for Gazebo and the camera itself is modeled after the parameters of the Intel Realsense.



# Setting up SLAM in Simulation

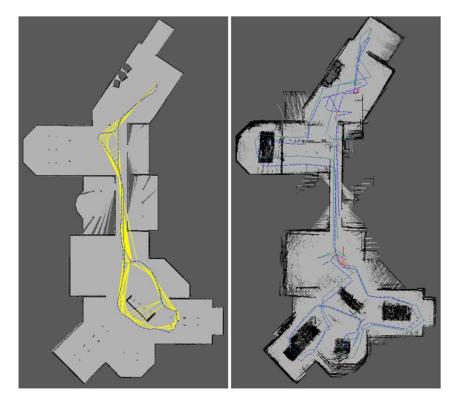
 The scenario that is used in this experiment is the Office world that is available from Clearpath Robotics.

 This is, as the name suggests, an office scenario that contains enclosed rooms with furniture & equipment



### Results

- The map on the left uses lidarodometry and therefore can estimate the map more accurately.
- The one on the right uses visual odometry and therefore, even though the geometry is captures, there are inconsistencies.



### Conclusion

- From the results of the experiments, it can be inferred that when in the context of an indoor mapping solution as proposed by the scope of this research work, RTAB-Map can be considered as a promising algorithm to perform the task with the required accuracy and reliability.
- From the experiments, the system has accurate localization as indicated by the low ATE (0.0292m) and rotational error.
- The measured boxes show a linear dimensional error below 15% and the average error of all the 3 dimensions falls below 10%. This indicates that the structure is also preserved accurately.
- Finally, the prototype puts this together in the form of a boxed solution, which can now be adapted onto standard robots running on ROS with minimal changes in terms of sensor topics and transformations.

# Scope of Expansion

- The first and foremost point for the expansion of this system should be by incorporating an accurate odometry scheme.
- Any method that improves the quality of depth improves the quality of mapping. Adding a highly accurate scheme for range estimation can be another front for expansion.
- Another front where this work can be further expanded is by incorporating semantic information into the sensory information.

# Thank you! ©

Questions?

### References

- [SK16] Bruno Siciliano and Oussama Khatib. Springer Handbook of Robotics. Springer Handbooks. Cham: Springer International Publishing, Jan. 2016, pp. 1–2227. isbn: 978-3-319-32550-7. doi: 10.1007/978-3-319-32552-1. url: <a href="https://link.springer.com/10.1007/978-3-319-32552-1">https://link.springer.com/10.1007/978-3-319-32552-1</a>
- [ESC14] Raul Mur-Artal and Juan D. Tardos. "ORB-SLAM2: An Open-Source SLAM System for Monocular, Stereo, and RGB-D Cameras." In: IEEE Transactions on Robotics 33.5 (Oct. 2017), pp. 1255–1262. issn: 1552-3098. doi: 10.1109/TRO.2017.2705103. url: http://ieeexplore.ieee.org/document/7946260/.
- [MT17] Raul Mur-Artal and Juan D. Tardos. "ORB-SLAM2: An Open-Source SLAM System for Monocular, Stereo, and RGB-D Cameras." In: IEEE Transactions on Robotics 33.5 (Oct. 2017), pp. 1255–1262. issn: 1552-3098. doi: 10.1109/TRO.2017.2705103. url: http://ieeexplore.ieee.org/document/7946260/
- [LM13] Mathieu Labbé and François Michaud. "Appearance-Based Loop Closure Detection for Online Large-Scale and Long-Term Operation." In: (2013). doi: 10.1109/TRO.2013.2242375.
- [LM19] Mathieu Labbé and François Michaud. "RTAB-Map as an open-source lidar and visual simultaneous localization and mapping library for large-scale and long-term online operation." In: Journal of Field Robotics 36.2 (Mar. 2019), pp. 416–446. issn: 15564967. doi: 10.1002/ROB.21831.