RoboBIM: A LiDAR-Based Autonomous Building Information Modelling System

Name: Lechen ZHANG

Student ID: 20031552

Supervisor: Adam Rushworth



Content

- Introduction and Background Information
- Aims and Objectives
- Hardware Design and Design Rationale
- Software Methodology
- Results and Discussion
- Conclusion
- Q&A

What is and Why BIM?



Information in BIM

'As-Built' BIM have significant practical meaning in building quality evaluation, heritage building maintenance, and protection and building environment assessment.



'As-planned' BIM and 'As-built' BIM

3

Why Unsupervised BIM?



The current building process of the 'As-built/As-is' BIM model on Revit software (Thomson, 2016)

Motivation of replacing this monotonous, repetitive, time-consuming, and errorprone manual modelling process is raising in recent years.

Aims and Objectives

In this project, an autonomous scanning mobile platform and a supporting software systems is proposed to achieve Level 3 (Adán et al., 2020) autonomous BIM Process.

Class of Autonomou s	Description
Level 1	Automatic data acquisition of the building's as-is state
Level 2	Simple geometric building model
Level 3	Recognition and labelling of primary structural elements (SEs) of the building
Level 4	Recognition of openings such as window and doors within SEs of the building
Level 5	Recognition of small building service components (BSCs) on SEs

Chart 1. Level of Autonomous BIM

Objectives

- (1) Complete design and setup of a robust mobile scanning platform
- (2) Make cost-effective decision in selection of hardware components and custom parts manufacturing to ensure expenditure does not exceed the ¥85,000 budget
- (3) Achieve the indoor robot navigation for autonomous building scanning
- (4) Getting a reconstructed building geometry model with centimeter level error
- (5) Successfully segment the building by main structures, including unobstructed walls, ceiling and floor. For the walls, flat walls larger than 1.5m*1.5m is considered.



Hardware Design and Design Rationale

- Selection of Sensors and Mobile Platform
- Main Structure Design and Setup
- Power System Design and Setup
- Signal Communication Design and Wiring

Description		Differential Wheeled Robots	Ackermann steering wheeled robot	Mecanum wheels robot	Multi-leg robot	Drone	Description		Wheel-Tech Ackermann Platform	Segway RMP401 PRO	Agilex Hunte
Criteria	Weight	Datum Model	Model 1	Model 2	Model 3	Model 4	Criteria	Weight	Datum Model	Model 1	Model 2
Cost	3	0	-			0	Cost	2	0		
Difficulty of Control	4	0	-				Protection Rate	4	0	+ + + +	+ +
Resource Available	3	0	0			-	Off-Road	Α	0		
Load Capacity	2	0	0	0	-		Ability	4	0	+ + +	+ +
Noise	1	0	0	-			Flexibility	4	0	+	
Durable	4	0	0	-							
Adaptability							Payload	1	0	+ + +	+ + + -
to Environment	5	0	+ +		+ + + +	+ + + +	Range	3	0	0	-
Safety	5	0	0	0	0		+		0	35	20
+		0	10	0	25	25	0		17	1	0
0		17	10	2	0	3			0	-1	_17
-		0	-7	-35	-40	-47			0	-4	-17
Net Score		0	3	-33	-15	-22	Net So	ore	0	31	3

Chart 2. Selection for Platform Type

Chart 3. Selection for Detailed Model of Ackermann-Drive Platform

Decision Matrix for Mobile Platform Selection

nter 2.0

(a) Front of the vehicle

(b) Back of the vehicle

Velodyne VLP-32C LiDAR

Huge FOV Mechaninical LiDAR Spinning Scan Pattern Repetitive Scanning

Livox MID-70 LiDAR

Small FOV Semi-Solid-State LiDAR Spiraling Flower Scan Pattern Non-Repetitive Scanning

Light Detection and Ranging (LiDAR) Sensor Selected for This Project

Power System Initial Design

10

Power System Final Setup

Signal Communication System

Software Methodology

The software environment of all the high-level control will base on NVIDIA Jetpack 4.6.1, a costumed Ubuntu 18.04 Linux system.

Robot Operating System (ROS)(Quigley et al., 2009) will be used for the communication of software and hardware. The achievement of the mapping, navigation and 3D reconstruction will highly rely on the function of ROS.

Robot Operating System

Setup of Simulation Environment

(a)

WHY LOAM?

LiDAR Odometry and Mapping (Zhang and Singh, 2014)

WHY Ground-Optimized LOAM?

Light-Weight Ground-Optimized LOAM (Shan and Englot, 2018)

KITTI Dataset is used for benchmarking (Geiger et al., 2012)

Reconstruction System Setup

LOAM and Poisson Reconstruction (Kazhdan et al., 2006) in IAMET 2F, UNNC

Segmentation System

LiDAR Intensity-Based Segmentation

Result and Discussion

Position	Truth Data (m)	Reconstruct ed Data (m)	Error (cm)	Percentage Error (%)
1	2.435	2.425	-1.0	0.411
2	19.515	19.478	-3.7	0.190
3	2.294	2.287	-0.7	0.305
4	12.703	12.684	-1.9	0.149

Chart 6. Translational Error Estimation

Result and Discussion

High Quality Scanning from Livox LiDAR

Degenerated Issue of Livox LiDAR

Future Improvements: Multi-Sensor Fusion

Conclusion

Reference List

ADáN, A., QUINTANA, B., PRIETO, S. A. & BOSCHé, F. 2020. An autonomous robotic platform for automatic extraction of detailed semantic models of buildings. *Automation in Construction*, 109, 102963.

GEIGER, A., LENZ, P. & URTASUN, R. Are we ready for autonomous driving? the kitti vision benchmark suite. 2012 IEEE conference on computer vision and pattern recognition, 2012. IEEE, 3354-3361.

HESS, W., KOHLER, D., RAPP, H. & ANDOR, D. Real-time loop closure in 2D LIDAR SLAM. 2016 IEEE International Conference on Robotics and Automation (ICRA), 2016-05-01 2016. IEEE.

KAZHDAN, M., BOLITHO, M. & HOPPE, H. 2006. Poisson surface reconstruction. *Proceedings of the fourth Eurographics symposium on Geometry processing.* Cagliari, Sardinia, Italy: Eurographics Association.

MCNAUGHTON, M., URMSON, C., DOLAN, J. M. & LEE, J. Motion planning for autonomous driving with a conformal spatiotemporal lattice. 2011 IEEE International Conference on Robotics and Automation, 9-13 May 2011 2011. 4889-4895.

QUINLAN, S. & KHATIB, O. Elastic bands: connecting path planning and control. [1993] Proceedings IEEE International Conference on Robotics and Automation, 2-6 May 1993 1993. 802-807 vol.2.

SHAN, T. & ENGLOT, B. Lego-loam: Lightweight and ground-optimized lidar odometry and mapping on variable terrain. 2018 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS), 2018. IEEE, 4758-4765.

THOMSON, C. 2016. From Point Cloud to Building Information Model: Capturing and Processing Survey Data Towards Automation for High Quality 3D Models to Aid a BIM Process.

VIVET, D., GéROSSIER, F., CHECCHIN, P., TRASSOUDAINE, L. & CHAPUIS, R. 2013. Mobile Ground-Based Radar Sensor for Localization and Mapping: An Evaluation of two Approaches. *International Journal of Advanced Robotic Systems*, 10, 307.

Thanks For Listening Q&A