Semantic Map Augmentation for Robot Navigation: A Learning Approach based on Visual and Depth Data

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Introduction



"For the next level of robot intelligence and intuitive user interaction, maps need to extend beyond geometry and appearance they need to contain semantics."

Source: SemanticFusion: Dense 3D Semantic Mapping with Convolutional Neural Networks - ICRA 2017



Introduction - Motivation







Introduction - Our Goal

• Build **augmented maps** with semantic information





Introduction



Metric Map (SLAM)

Object Detection



Introduction



Augmented map with semantic classes



Methodology





VERab UFMG



VERab UF**M**G







VERab UF¹¹¹G





Pipeline





Pipeline - Metric Map (SLAM)





Pipeline - Metric Map (SLAM)



Grisettiyz, Giorgio, Cyrill Stachniss, and Wolfram Burgard. "Improving grid-based slam with rao-blackwellized particle filters by adaptive proposals and selective resampling." *Proceedings of the 2005 IEEE international conference on robotics and automation. IEEE, 2005.*



Pipeline: Object Detection





Pipeline: Object Detection

YOLO: You Only Look Once



Redmon et al. You Only Look Once: Unified, Real-Time Object Detection, CVPR 2016

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Redmon, Joseph, and Ali Farhadi. "YOLO9000: better, faster, stronger." arXiv preprint (2017).



Pipeline: Object Localization





Object Localization



Point cloud representation of input camera has one-to-one correspondence with RGB image pixels.



Pipeline: Object Localization



The object localization is done using the point cloud of the detected class.



Pipeline: Object Tracking



Purple dots represent observations of the same class. Multiple observations of the same class might indicate a single instance (i.e. the object) or multiple instances of that class.



Pipeline: Object Tracking

- Each new instance of an object is modeled with a different Kalman filter object
- We start modelling "door" objects:

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- Doors are important for navigation and scene representation.
- Simple geometric model (plane patch) and tracking:

$$\begin{cases} \mathbf{x}_{i}[k] = \mathbf{x}_{i}[k-1] + \tilde{\mathbf{w}}[k] \text{ and } \tilde{\mathbf{w}} \sim \mathcal{N}\left(\mathbf{0}_{3\times 1}, \mathbf{W}\right) \\ \mathbf{y}[k] = \mathbf{x}_{i}[k] + \tilde{\mathbf{z}}[k] \text{ and } \tilde{\mathbf{z}} \sim \mathcal{N}\left(\mathbf{0}_{3\times 1}, \mathbf{Z}\right) \end{cases}$$

Experiments and Results



Online Experiments: Test Platform



Kobuki Base, RGBD camera, 2D Lidar.



Online Datasets

- Recorded datasets: rosbag of several sequences
- Available for download at : https://www.verlab.dcc.ufmg.br/semantic-mapping-for-robotics/

Results

- 1. Augmented map with semantic classes
- 2. 3D rendering of robot view

Results

VERab

Top-view of constructed map. Doors are represented in green

Future Work

Future Work

Add classes and their 3D model

Future Work

Semantic SLAM

Project page and source code can be found at:

https://www.verlab.dcc.ufmg.br/semantic-mapping-for-robotics/

Thank you! Questions?

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