

Semantic Robot Programming for Goal-Directed Manipulation in Cluttered Scenes

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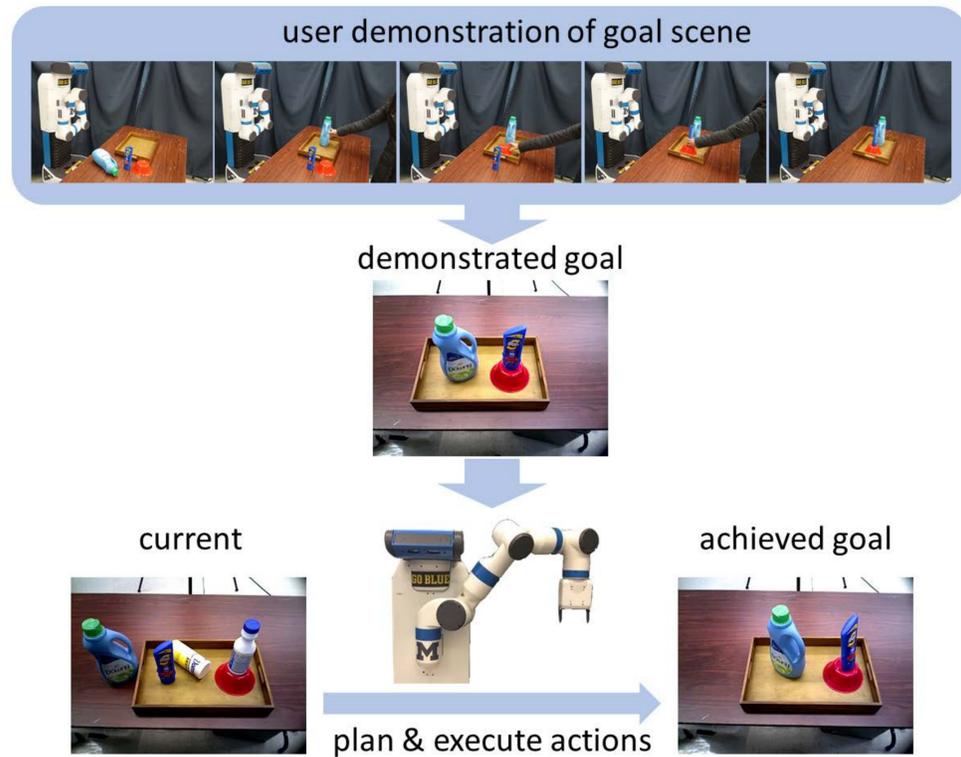
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Introduction



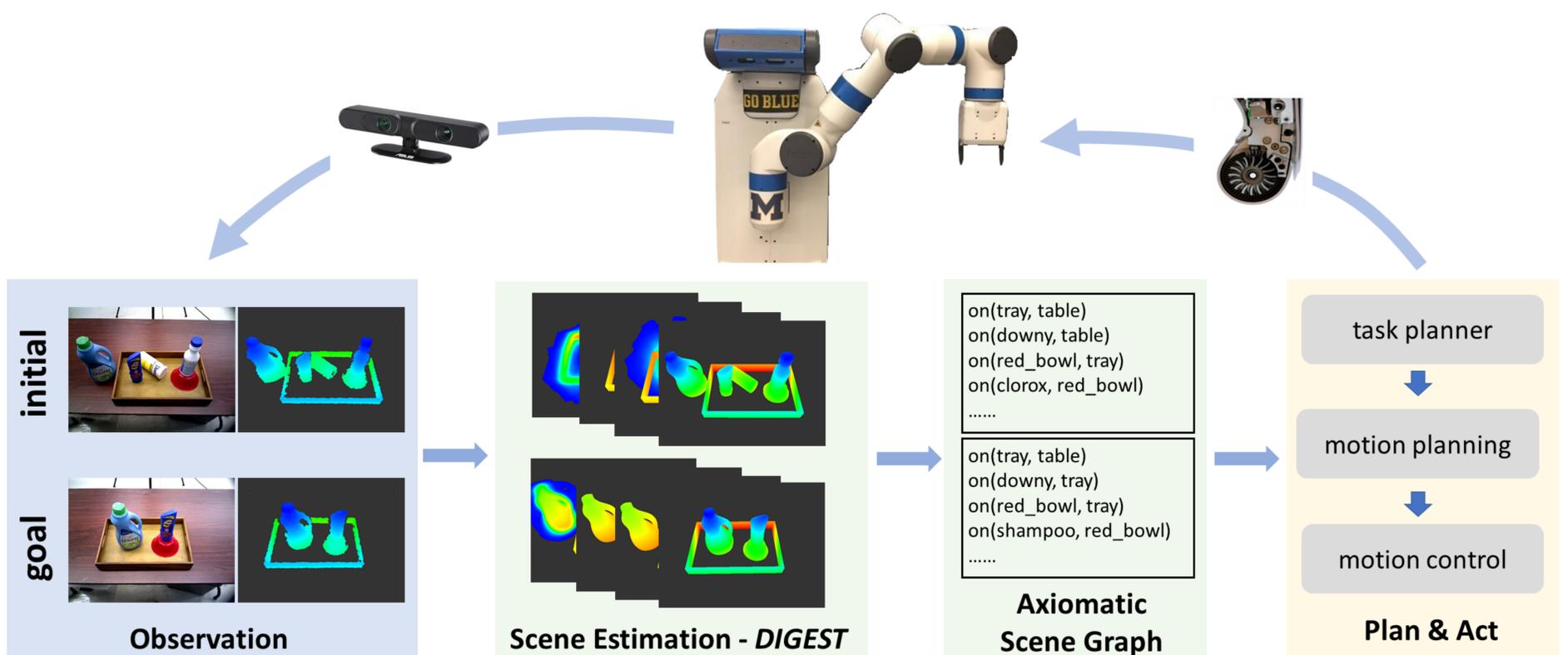
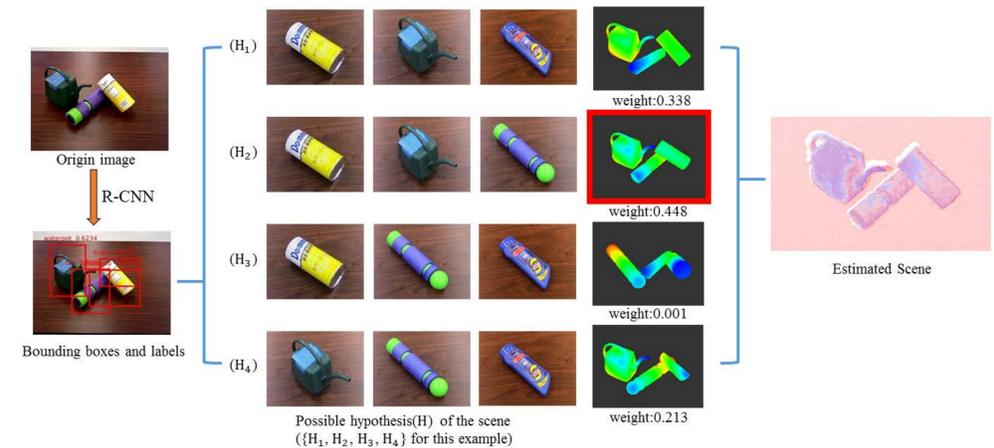
We present the **Semantic Robot Programming (SRP)** paradigm as a convergence of robot programming by demonstration and semantic mapping. In *SRP*, a user can declaratively program a robot manipulator by demonstrating a snapshot of their intended goal scene in workspace. The efficacy of *SRP* is demonstrated for the task of tray-setting with a Michigan Progress Fetch robot. Scene perception is evaluated on a public household occlusion dataset and our cluttered scene dataset.

SRP Overview

Given an RGB-D observation of the previously demonstrated goal scene and current scene, the robot parses each as an axiomatic scene graph, comprised of object poses and inter-object relations. Then the robot plans and executes a sequence of manipulation actions to rearrange objects to a satisfying goal scene. *SRP* assumes the number of objects present, and 3D mesh models for all objects.

Scene Estimation

Within *SRP*, we propose **Discriminatively-Informed Generative Estimation of Scenes and Transforms (DIGEST)** to estimate scenes from RGB-D observations. DIGEST refines object detections from R-CNN through generative 6 DOF pose estimation. Axioms representing the estimated scene graph are heuristically derived from the inferred object poses.



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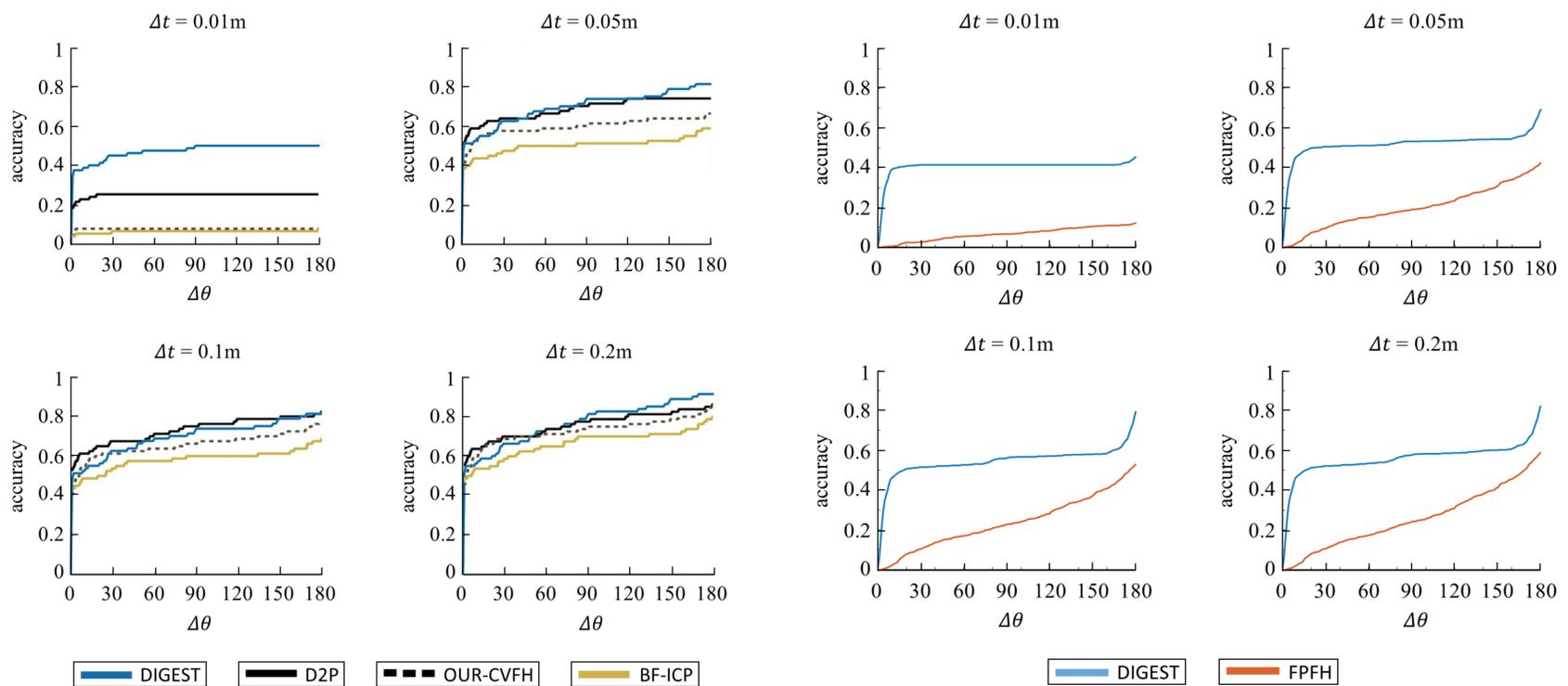
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Experiments

We evaluate the proposed *DIGEST* scene estimation method on a public household occlusion dataset^[1] and our cluttered scene dataset. *DIGEST* outperforms the state of the art method D2P^[2] on the household occlusion dataset, and outperforms FPFH^[3] on our cluttered scene dataset.



*y-axis: pose estimation accuracy; $\Delta\theta$: rotation error threshold; Δt : translation error threshold

Our *SRP* experiments are designed around a service robot scenario, where the robot needs to prepare a tray as demonstrated by the user in the goal scene. *SRP* is tested on scenes of 4 to 6 objects including the tray, with different inter-object relations. When faced with different initial scene configurations, *SRP* enables the robot to seamlessly adapt to reach the demonstrated goal.



References

- [1] Aitor Aldoma, Zoltan-Csaba Marton, Federico Tombari, Walter Wohlkinger, Christian Potthast, Bernhard Zeisl, Radu Bogdan Rusu, Suat Gedikli, and Markus Vincze. Point Cloud Library. *IEEE Robotics & Automation Magazine*, 1070(9932/12), 2012.
- [2] Narayanan, Venkatraman, and Maxim Likhachev. "Discriminatively-guided Deliberative Perception for Pose Estimation of Multiple 3D Object Instances." *RSS*. 2016.
- [3] R. B. Rusu, N. Blodow, and M. Beetz. Fast point feature histograms (fpfh) for 3d registration. In *Robotics and Automation. ICRA*. 2009.