

Objective

Use a depth camera and a turntable to create 3D representations of objects. Use the depth camera to generate point cloud data and a turntable (with position feedback) to spin the object and map all sides/features. The utility of this is that the full representation of the object could then be used for collision detection, grasp planning, path planning, etc, whereas with only a single-side representation, large clearances may be required to ensure no collisions are made with the unknown sides of the object, a grasp might not be possible, etc. ROS will be used as the framework, and the majority of the required code will be written in C++. Effort to tie in concepts from concurrent computer vision and sensor network courses will be made wherever appropriate, as well as relevant past course work (ME495, EECS469).

Process, Milestones, Timeline

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| 1. Pework: gather required info, order required hardware (turntable with feedback), determine object(s) to use, workspace setup, assess feasibility with Jarvis / prune project scope, etc | 1 week |
| 2. Depth camera correctly differentiates between object of interest and turntable/background | 2 weeks |
| 3. Determine and store 3D point representation of visible surfaces (static, no turntable rotation involved yet) | 2 weeks |
| 4. Turntable rotation added, dynamic surface addition working and capable of 3D point representation of more than just the visible surface | 4 weeks |
| 5. Markers added to the 3D representation to 'see' occluded sides/features of the object as it is spun, saved to object file as they are added | 1 week |
| 6. Implement any stretch goals, if time allows | |
| 7. Project demonstration, documentation, wrap-up | 1 week |

Stretch Goals

- write an algorithm to determine the optimal gripping configuration (required pose of the gripper relative to the object to best pick it up)
- *even stretchier*: then use Baxter and the turntable to execute the grasp
- remove turntable feedback and calculate the angle using the RGBD information
- *even stretchier*: make this work like SLAM (once features are connected, updates the positions based on most likely case - could potentially modify existing SLAM algorithm)
- Write an algorithm to simplify the point cloud representation of the object. For example, remove unnecessary/extra points from planar surfaces, edges, etc.
- incorporate depth camera measurement uncertainty to provide pseudosurfaces representing 95% confidence intervals?

Questions / Unknowns

- is it better to do average of points (i.e. spin the turntable 3 times and map the features from the average of the 3 spins)? Would this require a Bayesian filter?
- Which camera would be best for this purpose? Xtion? Kinect? Other?

- Any chance there will be resolution issues? What is the minimum distance between camera and object that the camera needs to detect?
- What is tolerance on depth measurements?
- Will distortion be a problem? Is it possible to calibrate to reduce this distortion?