

Object Pose Estimation in Robotics Using a Low-Cost RGB-D Camera

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Background

- Picking up an object in a structured and predefined environment is no match for today's industrial robots
- The task becomes more cumbersome when the objects are moved by e.g. a human operator
- With today's increasing demand for human-robot interaction and cooperation, the robot needs to be able to adapt to unstructured environments



Problem

- Can we estimate the pose of an object i.e. position and rotation using an RGB-D camera?
- How accurately can we measure the pose of the object?
- Can we transform the pose estimation into the robot's coordinate reference system?



Contents

- Basics
 - RGB-D Camera Introduction
 - Point Clouds and Point Cloud Library
- Project Description and Results
 - Constructing Point Cloud Models
 - Object Pose Estimation
 - Connecting to a Robot
- Future Work and Conclusion



RGB-D Camera Introduction

- RGB camera + Depth sensor → 2.5 D
- Time-of-Flight (laser, phase-shift)
- Passive triangulation
- Active triangulation (structured light)



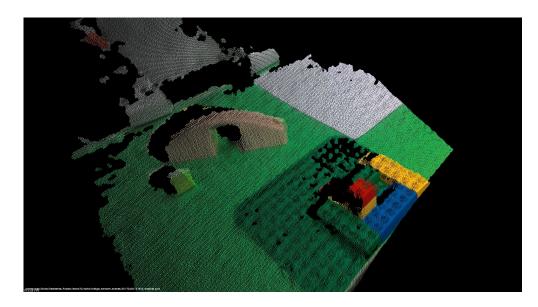
Intel RealSense SR300



- Small RGB-D camera using structured light
- Low-cost (around \$149)
- Uses Intel RealSense SDK with lots of examples
- Poor and incomplete documentation



Point Cloud



- Set of data points in 3D
- Can contain color information



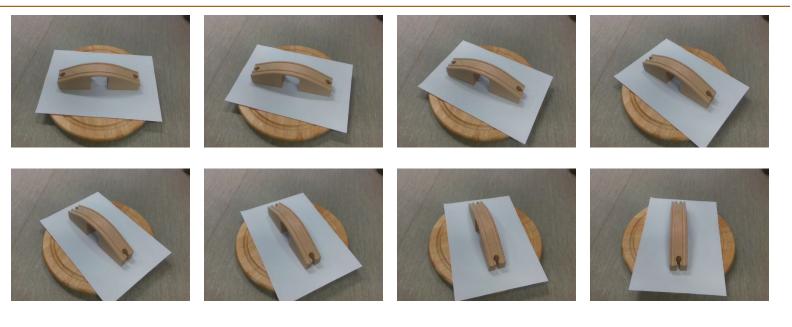
Point Cloud Library (PCL)

- Library for 2D/3D image and point cloud processing
- Large scale, open source and cross-platform
 - Well written documentation
 - Great tutorials
 - Supports many point cloud formats (PCD OBJ PLY)
- Written in C++
 - Some support for Visual Studio (2008 and 2010)



- We need a 3D model of the object for the pose estimation
- Achieved by segmentation and merging point clouds from different views







Segmentation

- RANSAC to find and remove plane
- Euclidean Cluster Extraction to extract point cloud of object





Random sample consensus (RANSAC)

- Mainly used as an outlier detector
- RANSAC to remove a plane:
 - Equation of a plane: ax+by+cz+d = 0
 - Inliers are points in a close proximity of the plane
 - Find the set of inliers to the plane and remove them

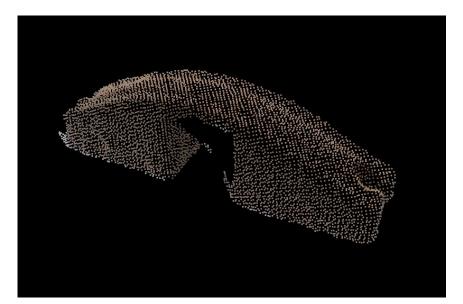


Euclidean Cluster Extraction

- Searches for the set of neighbors of a point that are within a sphere
- Uses a Kd-tree structure for finding the nearest neighbors









Pairwise registration

- Two consecutive segments have different rotation and translation.
- Pairwise registration tries to find the transformation between the segments
- Once the transformation has been found the point clouds are merged and smoothed.



Pairwise registration - Finding the transformation

- First a pose estimation algorithm is used to find a rough transformation using a heavily downsampled version of the segment (more on this later)
- Once a rough estimate of the transformation has been found an Iterative Closest Point (ICP) algorithm is used to get a better estimate using the full size of the sample set
- The point cloud in the second image is then transformed to match the point cloud in the first image

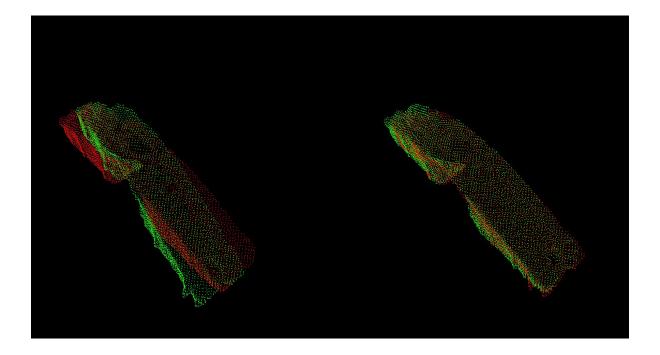
Iterative Closest Point (ICP)

- Used to minimize the distance between two point clouds.
- Achieved by estimating a rotation *R* and translation *t*
- This is done by minimizing a cost function:

$$V(R,t) = \sum ||x_i^{(1)} - Rx_i^{(2)} - t||^2$$

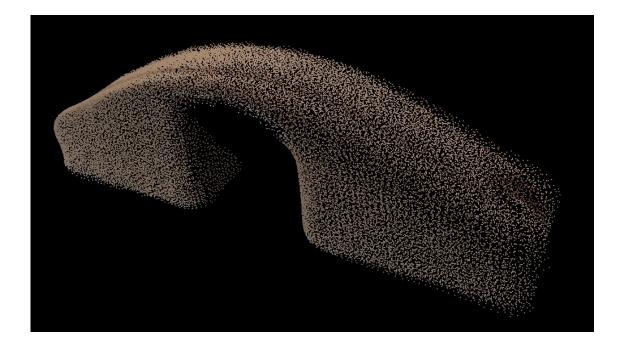
 Convergence is improved if the point clouds are initially close







Final Point Cloud Model





Final Point Cloud Model

- Extracted model is far from perfect
 - Hard to achieve a true 3D model of symmetric objects
- Alternatively:
 - Create and use a CAD model of the object
 - Use 3D-scanning software (price)
 - Reconstruct the 3D image by using the RGB images
 - Extract matching features and use triangulation

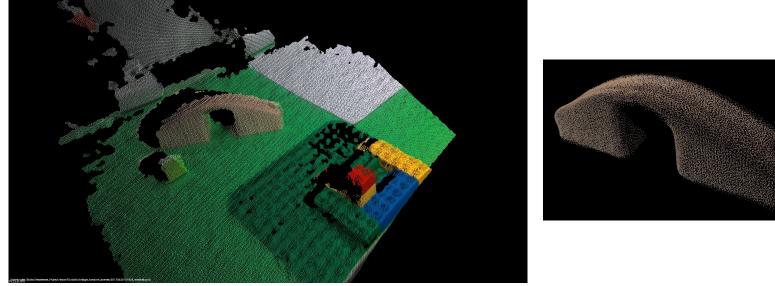


Object Pose Estimation

- We want to estimate 6-DoF
 - Position of object (3-DoF)
 - Rotation of object (3-DoF)

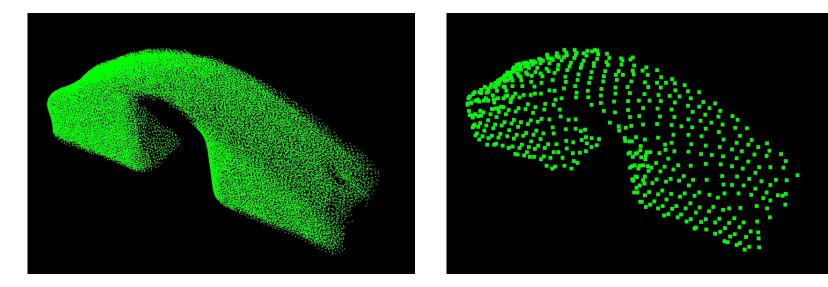


Pipeline - Input



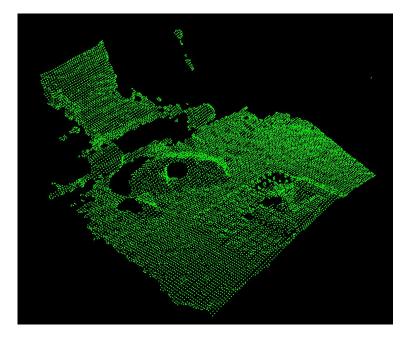


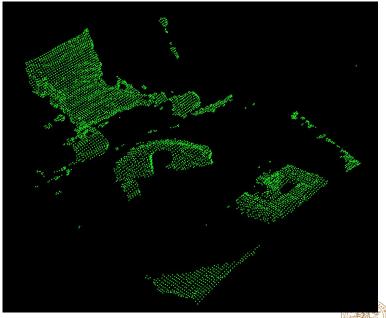
Pipeline - Downsample





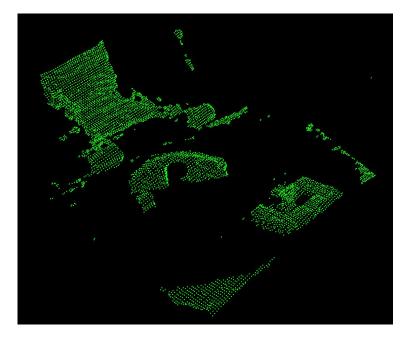
Pipeline - Remove largest plane

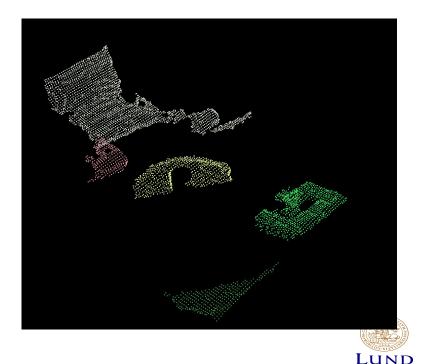




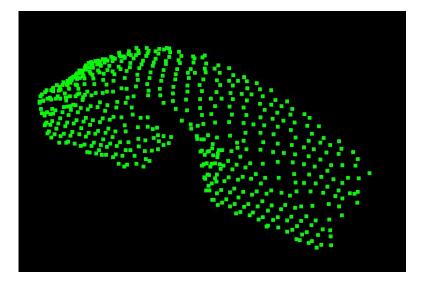


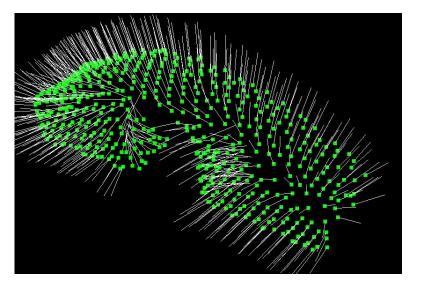
Pipeline - Extract clusters





Pipeline - Estimate normals







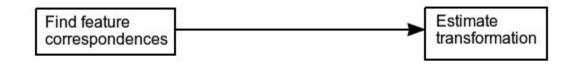
Pipeline - Estimate features

- Encode a points geometrical properties
- We use Fast Point Feature Histograms (FPFH)
 - Looks at the k-neighborhood of each point
 - Computes features based on direction of normals
- Many different point feature representations available
 - Viewpoint Feature Histogram (2010)
 - Color Point Pair Feature (2015)

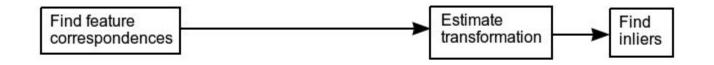


Find feature correspondences

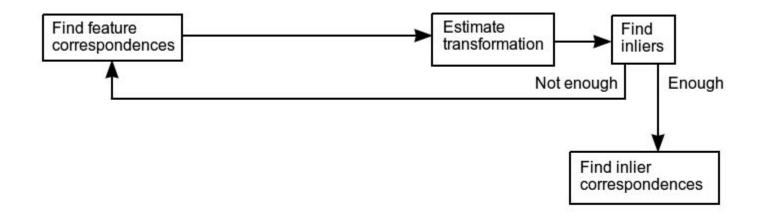




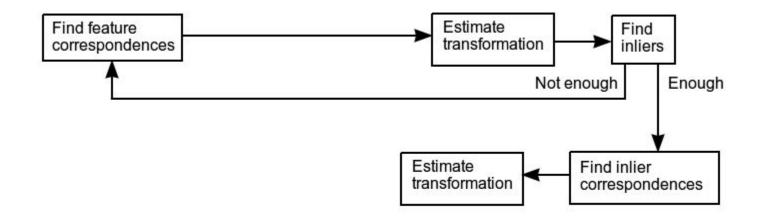




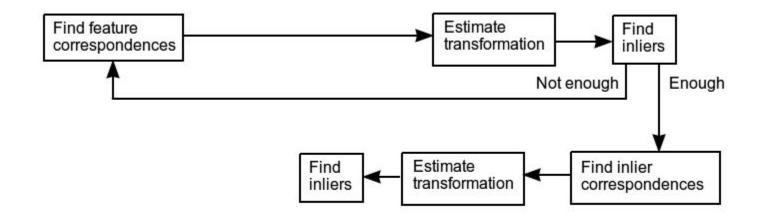




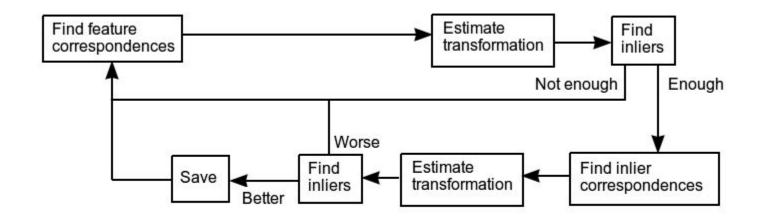




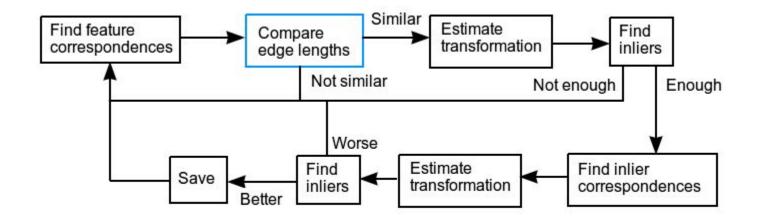






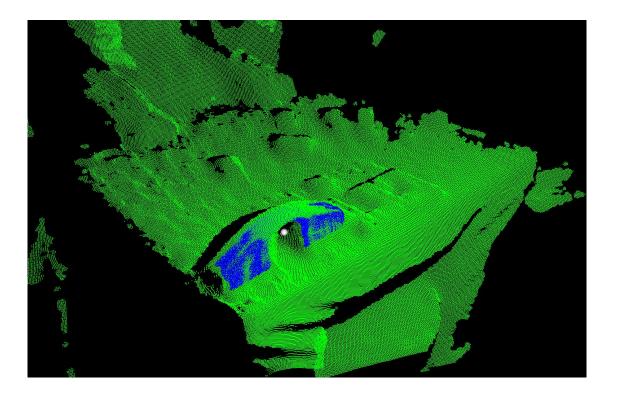






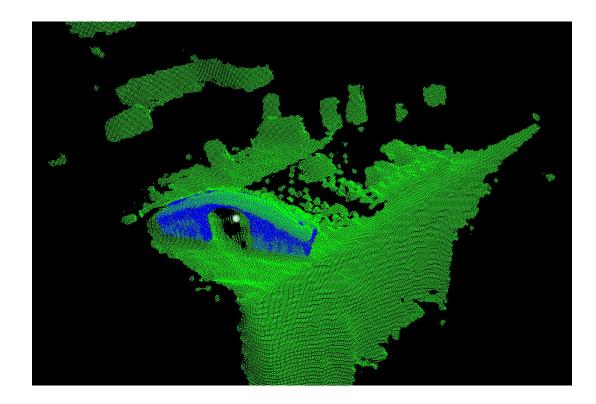


Results





Results





Connecting to a Robot

- Find transformation from camera to robot coordinate system
- Euclidean transformation
 - Need at least 3 points expressed in both coordinate systems
- Could be done automatically using pose estimation
- Could also directly estimate pose of robot if it is visible



Problem

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Future Work

- Autonomous robot-camera calibration
- Using RGB data in features
- Estimating pose for multiple identical objects
- Using multiple RGB-D cameras to avoid occlusions



Thank you for listening!

Questions?

