

Incremental Registration of RGB-D Images

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Introduction: The Problem

Problem description:

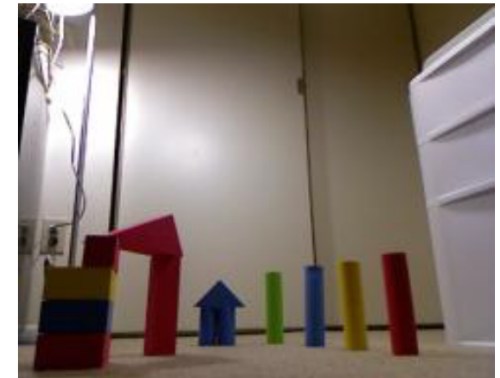
- Estimate the 6-DoF pose of a RGB-D camera in freehand motion
- High-frequency update described

Data input:

- Scans from an RGB-D camera (such as Kinect).
- We assume RGB and Depth images are already registered

Applications:

- A fast, robust pose estimation is useful for:
- Control
 - 3D Mapping & SLAM



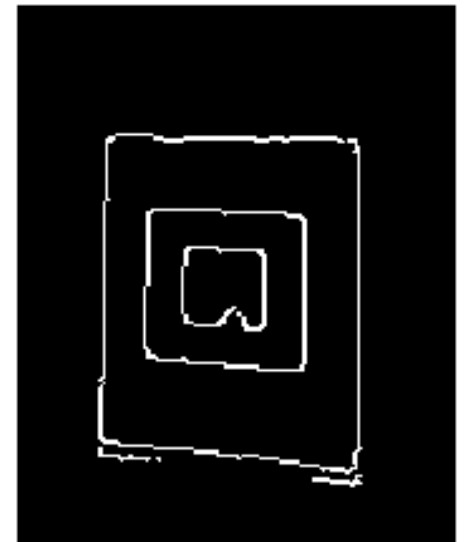
Introduction: The Approach

Overview of the approach:

- Hi-freq. loop – operates on sparse data
 - Detect edges
 - Filter edges
 - Classify edges
 - Perform Edge-ICP
- Lo-freq. loop – operates on dense data
 - Use output of hi-freq. loop as estimation for motion
 - Perform point-to-plane ICP

Edge detection

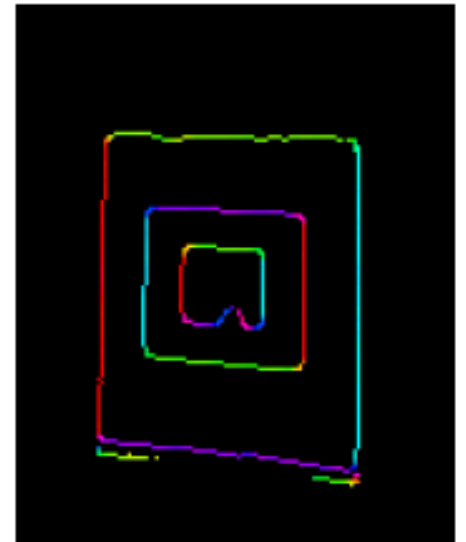
- 1) Convert RGB image to gray-scale
- 2) Perform Gaussian blur filtering to remove noise
- 3) Perform Canny edge detection to locate edges



Edge classification

Each pixel belonging to an edge is classified by the edge orientation

Orientation is computed using the gradients along the x- and y-dimensions.



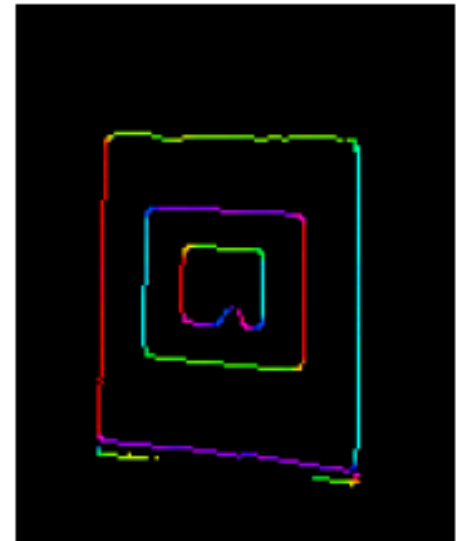
Edge classification

Each pixel belonging to an edge is classified by the edge orientation

Orientation is computed using the gradients along the x- and y-dimensions.

$$\theta = \text{atan}\left(\frac{dy}{dx}\right)$$

Where dy and dx are computed using a Sobel operator



Detection + classification

Example output for a typical image:

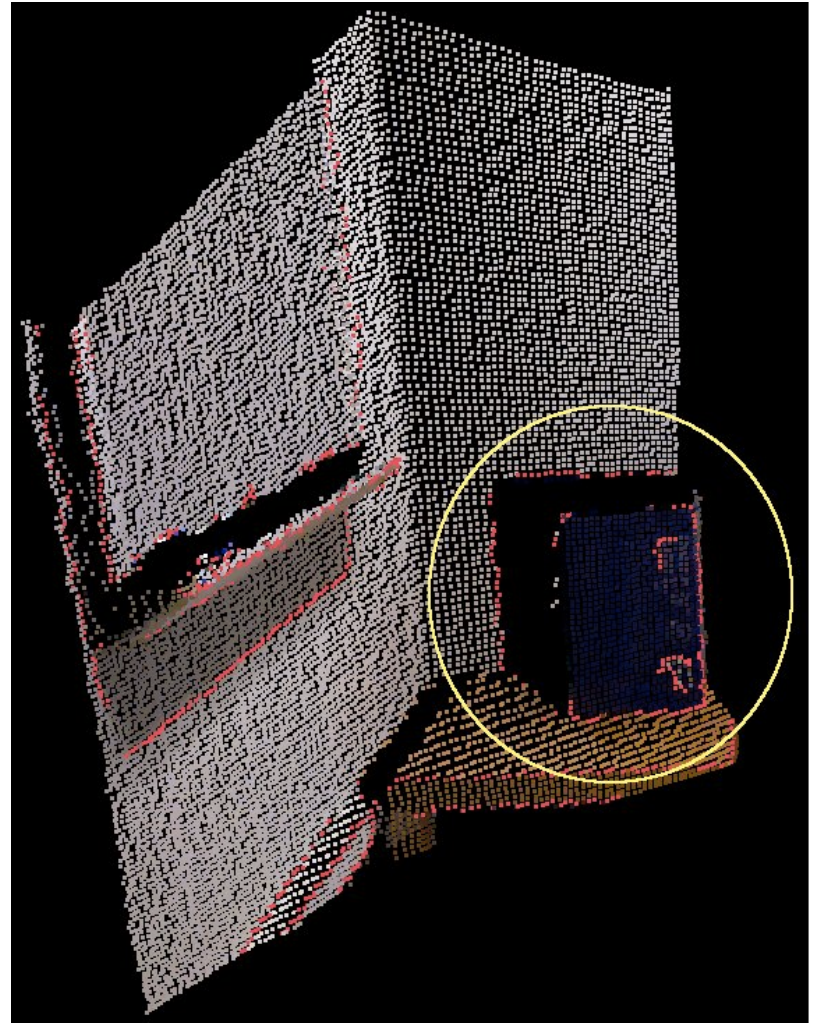


Edge filtering – why is it needed?

We create sparse point cloud from all the pixels belonging to edges (see image - pink pixels)

Problem: edge sometimes appears on the object background instead of object foreground (see image – area in circle)

However, we want edges which are *pose-invariant*.



Edge filtering algorithm

Problem:

Edge sometimes appears on the object background instead of object foreground

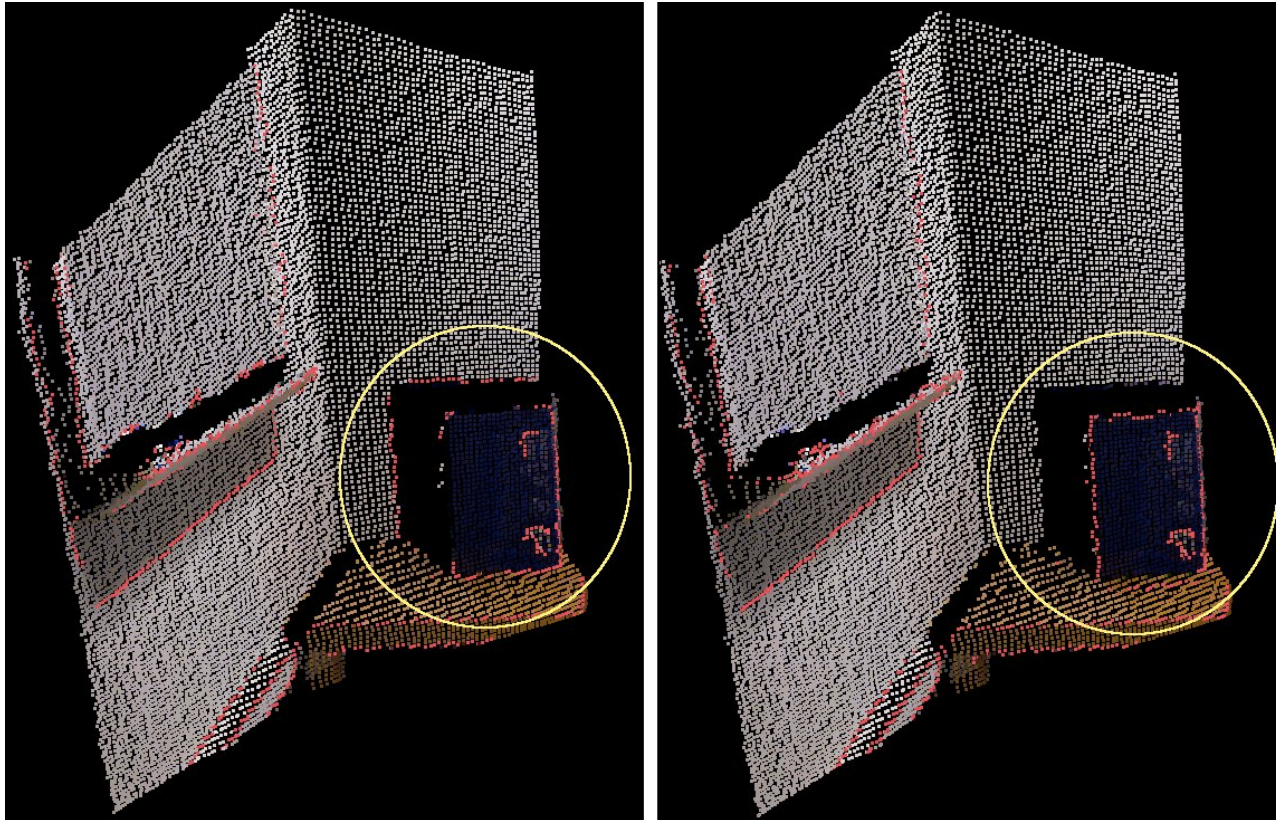
Solution:

Perform a local search around each edge pixel (window size of 3x3 or 5x5) in the Depth image

If there exists a significant jump from high depth value to low depth value, use the pixel with the lower depth value (closer to the camera).

Detection + classification

Example output of depth filtering:



Edges on book correctly determined, other edges remain the same.

Registration: high frequency loop

Edge-ICP algorithm:

Register sparse point clouds of edge points using ICP

Nearest neighbor correspondences are computed using Euclidean distance in 3D.

For each nearest neighbor found, we check if the θ values are the similar (within 30 degrees)

If not, we look for the next nearest neighbor.

This helps prune out false correspondences between edges

Registration: low frequency loop

The low-frequency loop runs parallel to the high-frequency loop in a separate thread

We use Generalized ICP (G-ICPP) to register the dense 3D data

Normally, GICP is slower. We use the estimated transform from the high-freq. loop as input to GICP

Result: low-freq. loop refines results from high-freq. loop.

Results:

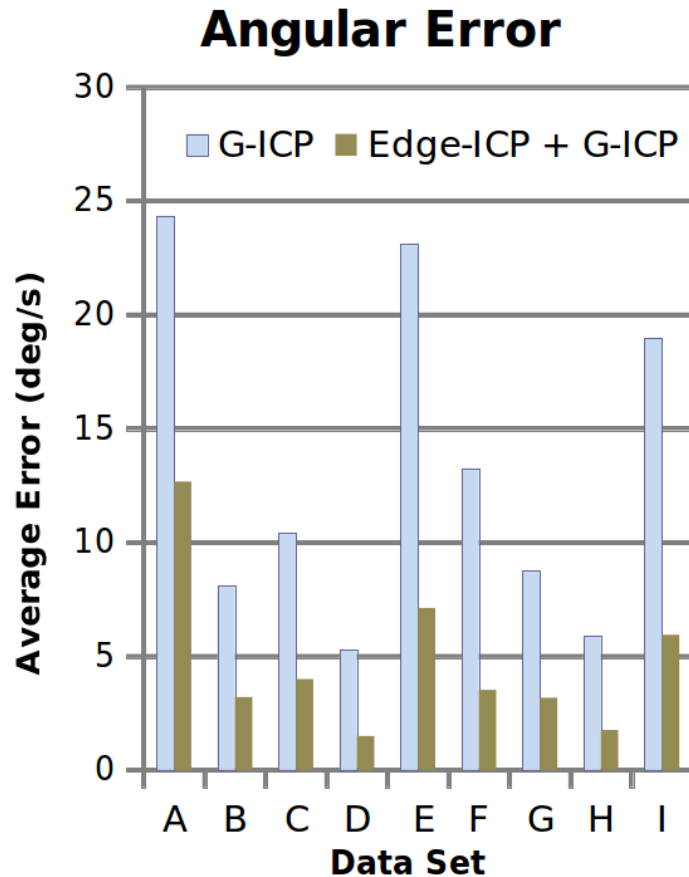
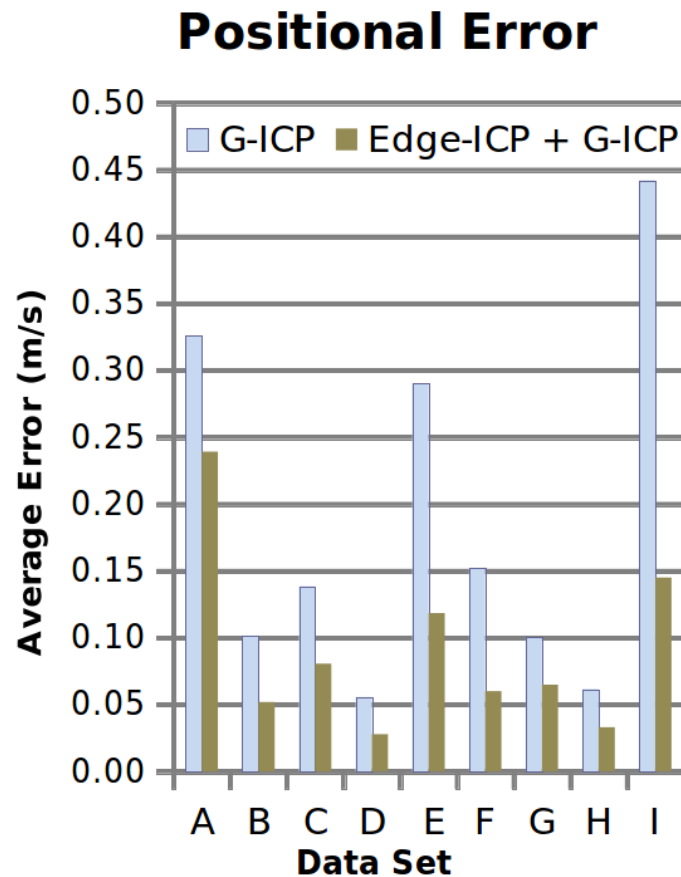
We use an RGB-D datasets with ground truth from a VICON motion-capture system.

Compare error from dual-loop approach(Edge-ICP + GICP) vs. single-loop (GICP) registration.

Measure average angular error and translational error per each frame.

Results:

How does the dual-loop approach compare vs. regular GICP?



Current work:

1) We are considering variety of features for the high-frequency loop:

- Canny edges
- ORB
- SURF

Evaluation of which offers better tradeoff between speed and accuracy

2) Incorporating a full SLAM system which can deal with loop closure

3) Open-source release, targeted for robotics applications.

Thank you!