Amazon Picking Challenge Stowage Task

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Introduction

- Designed robotic system for stocking warehouse shelves
- Task
 - Retrieve items out of an unstructured bin
 - Identification of each item
 - Place item onto the shelf without damage
- Competing in the 2016 Amazon Picking Challenge





System Overview



System Demo: Part 1



Bin Actuation

- Linear actuator tilts bin to increase accessible space
- Robot is able to pick up items against the walls



Motion Planning

- Movelt! software package manages arm kinematics and path planning
- OMPL motion planner implements RRT*Connect algorithm to find optimal, minimal distance paths
- Base, end effector, and shelf modeled as collision objects



Suction Grasping

- High flow low pressure vacuum system
 - Shop-Vac impeller provides 200 CFM and 40 kPA
- Custom suction cup mounted to UR5 wrist
- Capable of acquiring 36 / 38 items from the 2016 APC item list



Grasp Planning

1) Region Growing Segmentation

- Select points with minimum curvature values as cluster centers
- Find nearest neighbors of the selected points to generate clusters

2) Clusters are scored based on...

- Maximum number of points
- Height of each cluster
- Area of the horizontal surface of each cluster (x-y axis)
- Direction of normal



Item Identification

- Capture kinect2 RGBD data
- Mask image based on depth data
- Segment image using SLIC algorithm
- Classify superpixels using Caffe and Alexnet CNN architecture
- Make prediction based on



Dataset Generation

- Automatic image capture using actuated turntable
- HSV color based segmentation
- Convex hull approximation
- Collected 100 images for all 38 items





Item Segmentation

- Fixed location of the endeffector above the ID Kinect.
- Depth based thresholding
- Geometries of all the objects are known, horizontal thresholding.





Image Segmentation and Identification

- Segment image using SLIC superpixel algorithm
- AlexNet CNN is used for item identification
 - Trained for 8 hours on over 400,000 segments using Nvidia Titan GPU
 - Predictions are computed in < 1 second
- 38 predictions are generated for every super pixel



Global Prediction

- Average predictions for each class
- Identify item that has the highest confidence prediction
- Renormalize predictions
- Repeat until all 12 items are identified

 75% accuracy on over 10,000 simulated permutations of acquired images



System Demo: Part 2



Future Work



2 months to improve system before the competition at Robocup on June 31st

Grasping from bin

- Use improved clustering to improve first pass pickup rate
- **Item Identification**
 - Further supplement training data with real images
 - Place items in 'confusion set' in the same bin to reduce the chance of misassociations
 - Refine global prediction algorithm to maximize accuracy

Item placement on shelf

- Develop pose estimation method for large items
- Develop place strategy for large items of known pose

Questions

[1] Amazon picking challenge rules:

http://amazonpickingchallenge.org/APC 2016 Official Rules.pdf

[2] Amazon picking challenge submission video:

https://youtu.be/oGq05wN7mmg

[3] PCL Region Growing Segmentation:

http://pointclouds.org/documentation/tutorials/region_growing_rgb_segme_ntation.php

[4] Movelt: http://moveit.ros.org/

[5] Open Motion Planning Library: <u>http://ompl.kavrakilab.org/</u>

[6] AlexNet: <u>http://papers.nips.cc/paper/4824-imagenet-classification-with-</u> deep-convolutional-neural-networks.pdf

[7] Caffe Deep Learning: <u>http://caffe.berkeleyvision.org/</u>

[8] Berkeley APC Dataset: <u>http://rll.berkeley.edu/amazon_picking_challenge/</u>