

Given : Multiple objects are randomly arranged in a heap

Objective : Sequentially grasp and transport each into a packing bin

Challenges :

- Reliable Robotic Grasping is challenging due to imprecision in sensing and actuation.
- Difficult to infer object shapes from point clouds due to sensor noise, obstructions and occlusions when objects are placed in heaps
- For deep learning model, time cost of collecting physical data too high

Ideas:

- We model bin picking with Partially Observable Markov Decision process : States of heaps, point cloud observations, rewards
- Train on synthetic datasets of grasps and point clouds
- It is possible to grasp a diverse set of objects from clutter using Deep CNNs.
- Consider modeling uncertainty during dataset generation to learn a policy for rapid bin picking from a single view point
- Modelling bin picking as a sequence of 3D objects in heap with noisy point cloud observations using POMDP
- Deep neural networks trained on large datasets can predict grasp positions to a very high accuracy
- DexNet - Dataset of 6.7 million point cloud objects
- GQCNNs- Grasp Quality Convolutional Neural Networks : Rapidly predicts the success of grasps from depth images
- Grasps are specified as planar position, angle and depth of a gripper relative to and RGB-D Sensor

Assumptions:

- Camera is mounted on a stationary position to capture the top view of the heap

- Gripper motion can occlude camera view
- Point cloud of one object selected as X for CNN
- Gripping position selected as y for CNN
- We perform 3D Point Cloud Segmentation

Process :

Generating the synthetic dataset for heap, Tools : NDDS, Dex-Net

- 1.3D CAD Models generate
- 2.Sampling from 3D CAD Models : NDDS / Dex-Net
- 3.Generate heaps of models using Dynamic simulation of dropping action
- 4.Heap state Target Specification, Tools : NDDS, Dex-Net
- 5.3D Point cloud
- 6.3D Point cloud segmentation
- 7.Specify target object to pick from the heap- Grasp Generation, Tools : Dex-Net, Isaac-Sim, iGibson, Gazebo
- 8.Define a trajectory that reaches the object without occlusions
- 9.Gripping action so as not to damage it
- 10.Policy between Gripping action and point cloud
- 11.Reward if successful transfer of object.
- 12.Choose the heap state/ point of object as X and gripping position as y for a CNN only for which reward is 1. Feeding the heap and grasp dataset to a CNN, Tools : Dex-Net (GQCNN used)
- 13.Feed to CNN
- 14.Predict trajectories for new heaps
- 15.Simulate the bin picking process, Tools : Gazebo, Isaac-Sim, iGibson, Neurorobotics