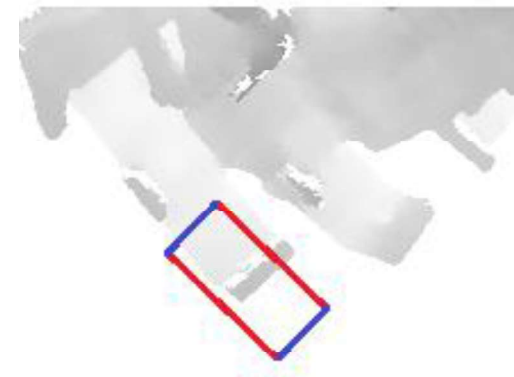
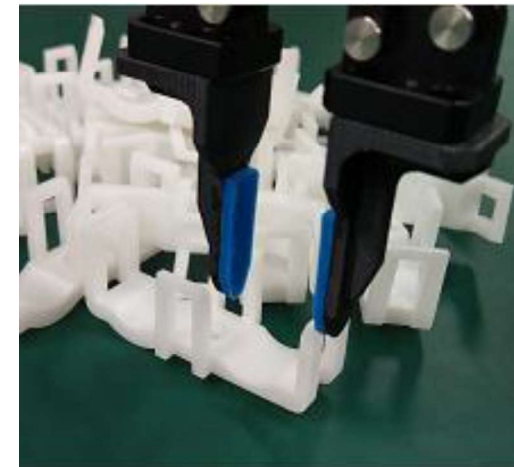


OUTLINES

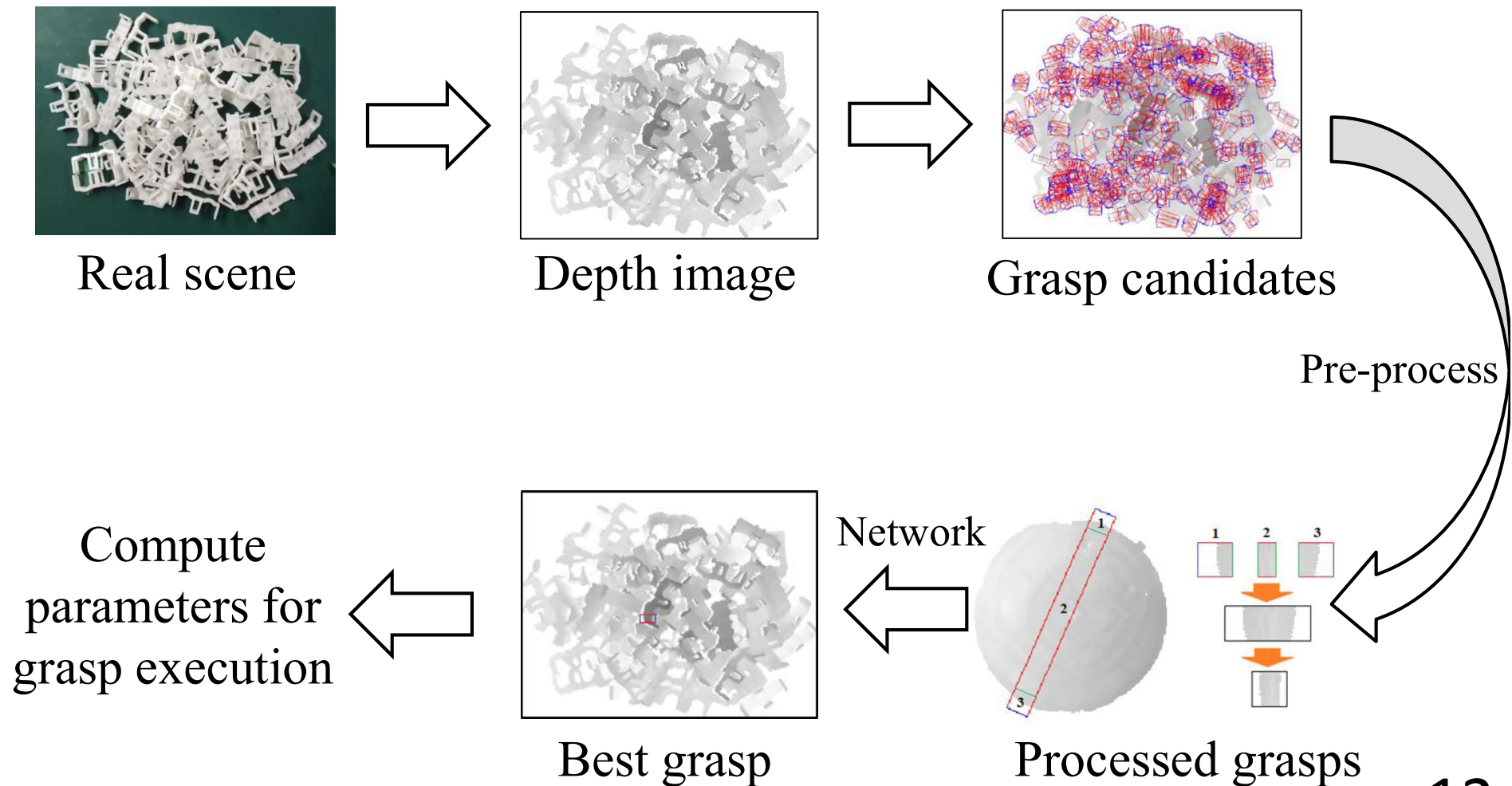
1. Research background
2. Research objectives
3. Related works
- 4. Grasp detection**
5. Pose estimation
6. Conclusion and future work

4.1 Rectangle representation of grasp

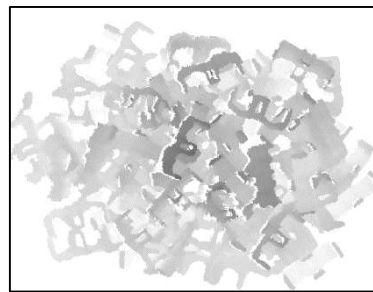
- Use rectangle to represent the grasp of a parallel-jaw gripper [10].
 - Blue lines indicate the layout of fingers
 - Length of red line indicates the open width
- Required parameters for grasp of a parallel-jaw gripper:
 - Intrinsic gripper finger size d .
 - Position of gripper center x, y, z .
 - Open width of the gripper w .
 - In-plane rotation angle θ .



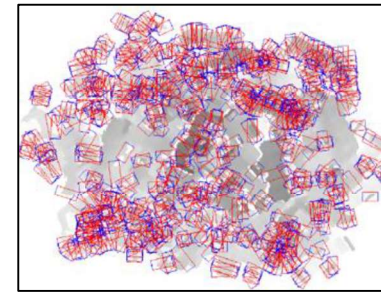
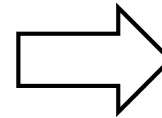
4.2 Algorithm pipeline



4.3 Generation of grasp candidates



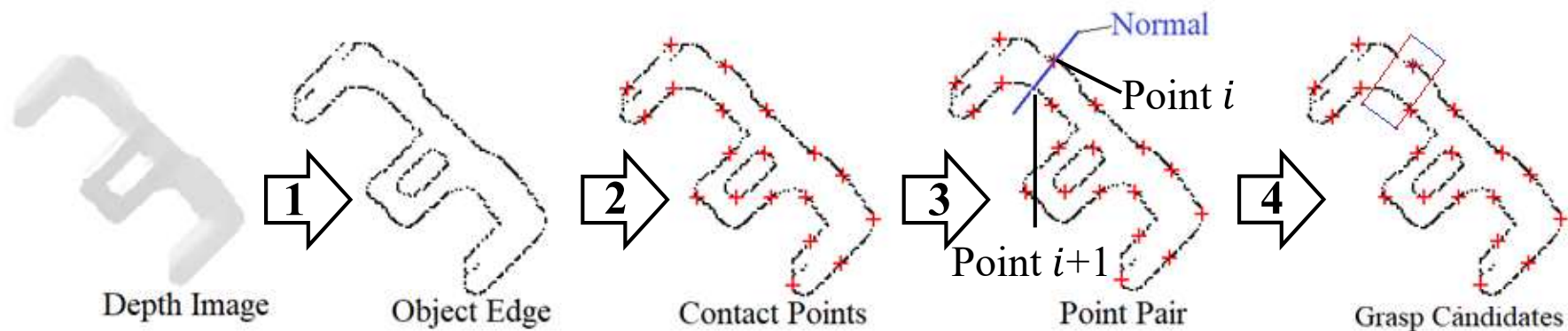
Depth image



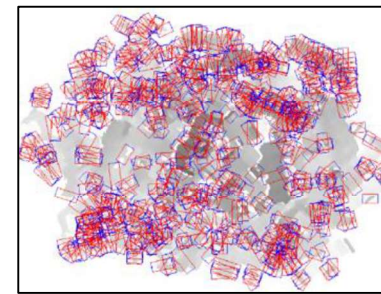
Grasp candidates

4.3 Generation of grasp candidates

- Generate grasps by contact points
 1. Obtain object edge from depth image.
 2. Find the contact points on object edges.
 3. For every contact point i , find the normal and the intersection of normal and edge as points $i + 1$.
 4. Generate a grasp candidate using two contact points.

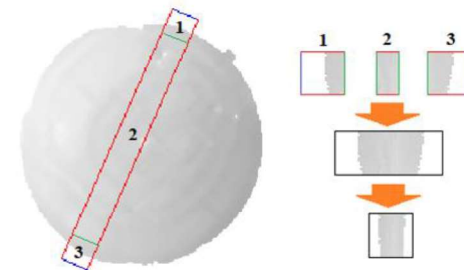


4.4 Pre-processing of grasp rectangle



Grasp candidates

Pre-process

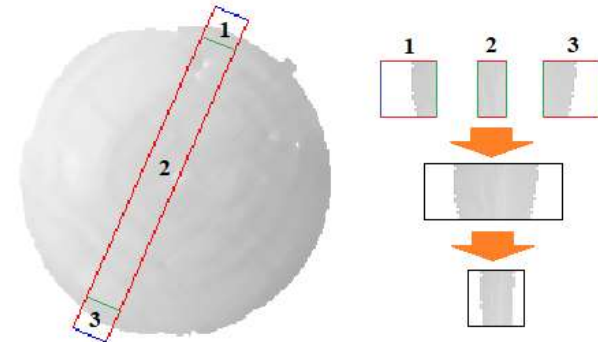


Processed grasps

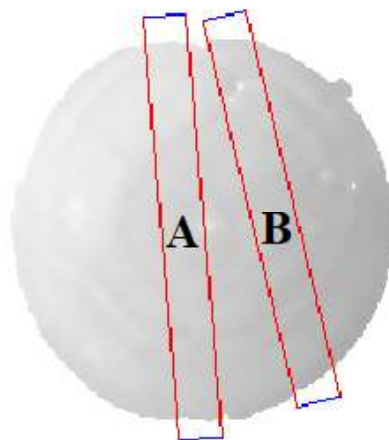
4.4 Pre-processing of grasp rectangle







- Process the input image to the same size.
 1. Separate the rectangle into 3 parts.
 2. Keep the size of part 1&3, and compress part 2
 3. Combine 3 parts and resize to the target size.



- Compared with simply resizing, the proposed method provides appropriate grasp rectangle and scores.



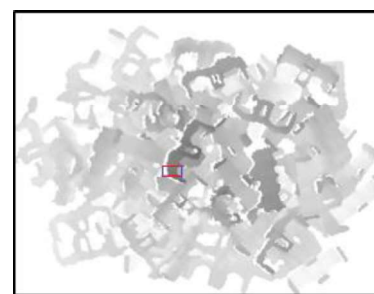
Grasp Scores Using Different Data Processing Methods

Grasp No.	A	B
Simply Resizing	 7.975	 8.709
Our Method	 8.079	 5.066

Grasp scores of simply resizing and proposed method

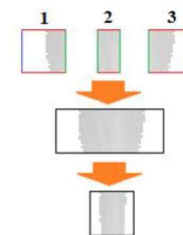
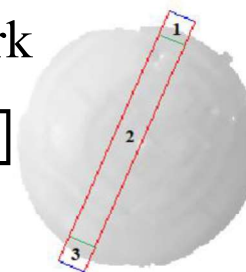
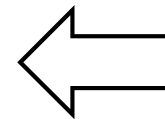
16

4.5 Network structure



Best grasp

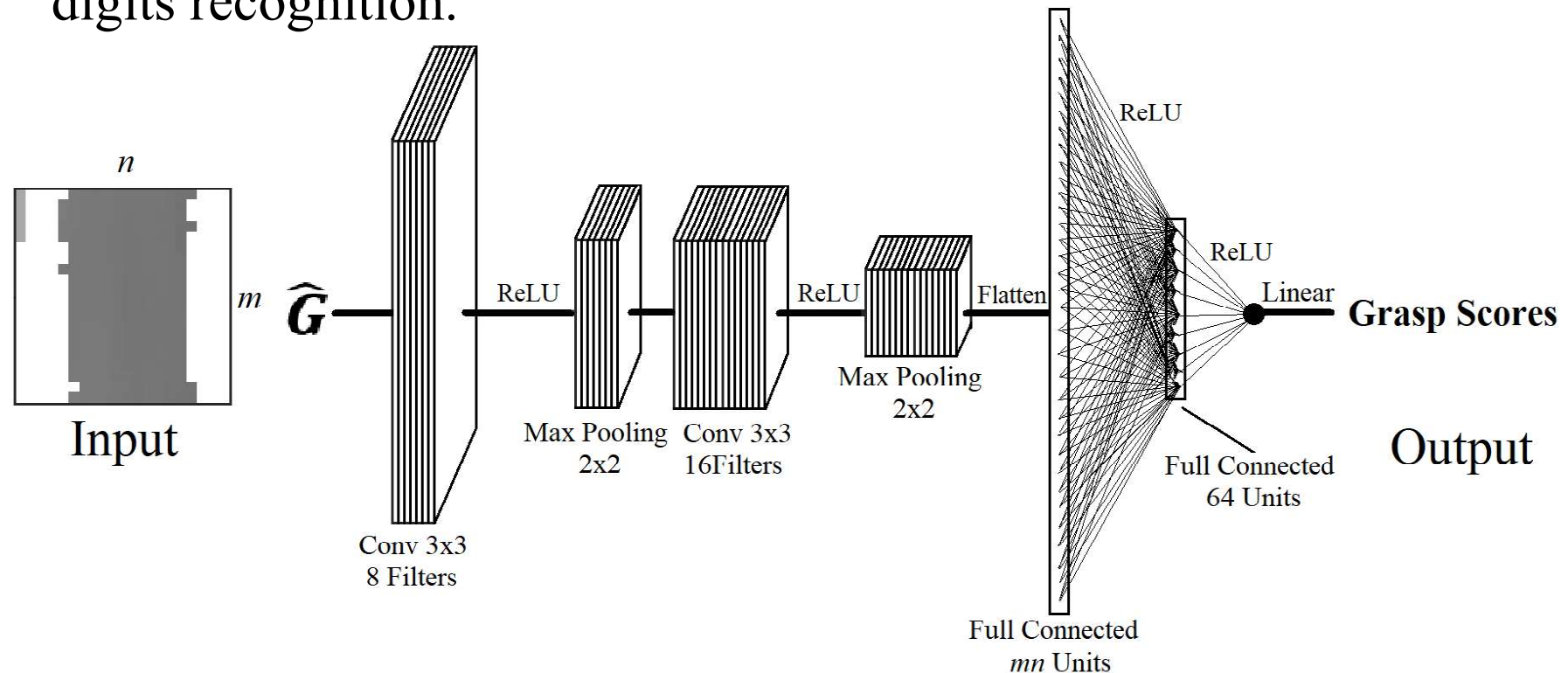
Network



Processed grasps

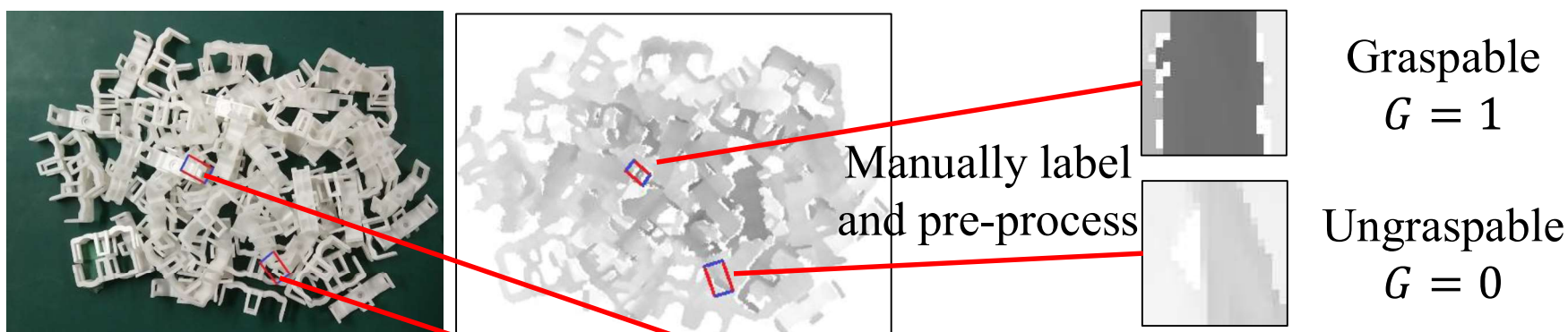
4.5 Network structure

- The structure is similar with LeNet-5 [8], we change the number of filters and layers, and use only 400 perceptrons. LeNet-5 is a very classical convolutional neural network used for handwritten digits recognition.



4.6 Training process

- For every type of object, prepare 2000 training data including 1000 graspable rectangles and 1000 ungraspable rectangles.



- Loss function:

$$L = -\frac{1}{N} \sum_1^N G \ln \sigma(\hat{y}) + (1 - G) \ln(1 - \sigma(\hat{y}))$$

\hat{y} is predicted scores by network, $\sigma(\hat{y}) = \frac{1}{1+e^{-\hat{y}}}$ [12]

Hyper-parameters:

- Data size: 20×20
- Dataset size: 2000
- Batch size: 32
- Dropout: 0.5
- Optimizer: RMSProp
- Learning rate: 0.001