## **OBJECT DETECTION AND GRASP PLANNING FOR AUTONOMOUS INTERVENTION** MARINE & MARITIME

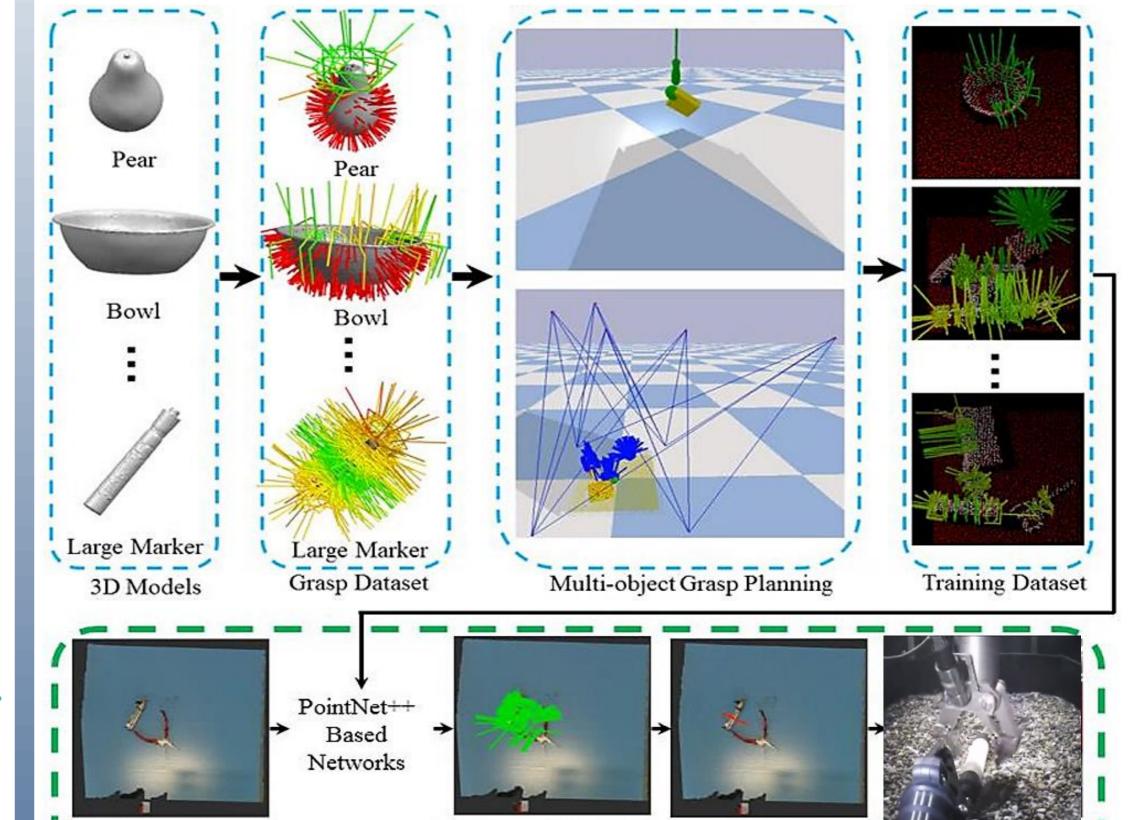
**Md Ether Deowan** 

Co-funded by the **Erasmus+ Programme** of the European Union



## **CONTEXT AND MOTIVATION**

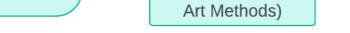




Reduces human intervention, increases efficiency and safety. Stereo camera **Revolutionizes underwater** operations with precise intervention. Integrating deep learning for object detection with state-of-the-art grasp planning algorithms. Fig1: GIRONA 500 I-AUV Capture **Object Detection** Still Image (Deep Learning) Data Acquisition Sequense (Stereo Camera)

INTELLIGENT

ROBOTICS



**Direct Object Detection** 

in Point Cloud

Propose Grasping

Poses (State of the

Segment Object

in Point Cloud

Choose

Segmentation

Strategy

Fig 2: Objective Flow Diagram.

Capture

Point Cloud

End

## **STATE OF THE ART SOLUTION**

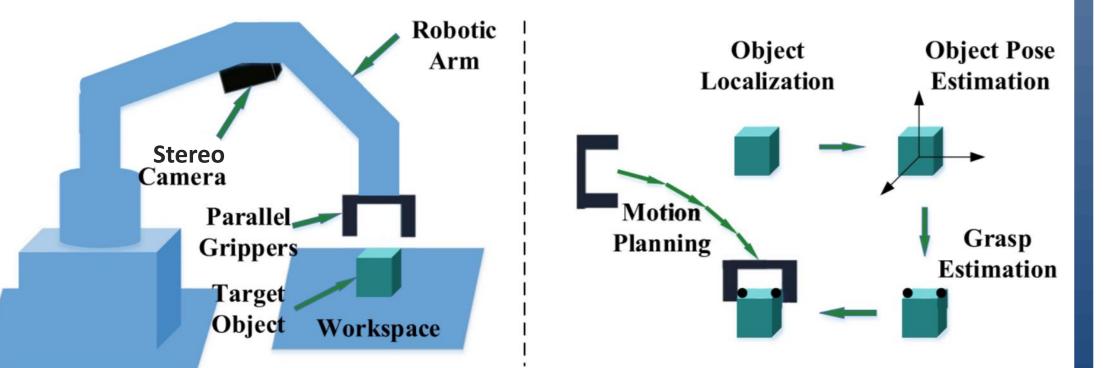
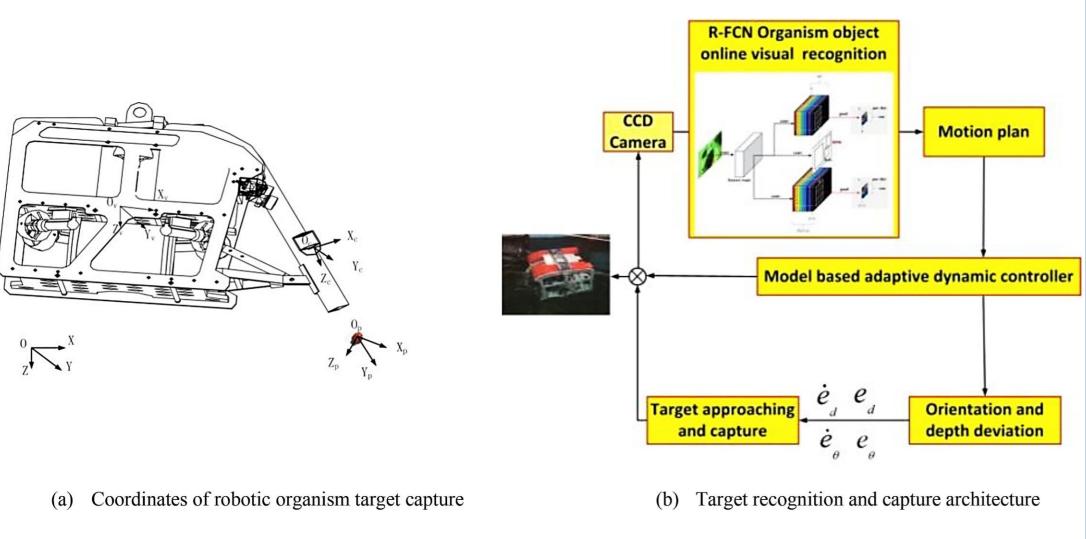
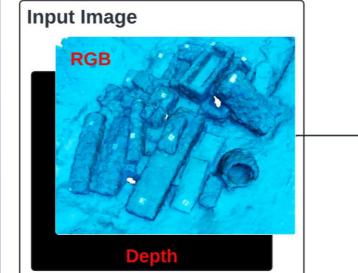


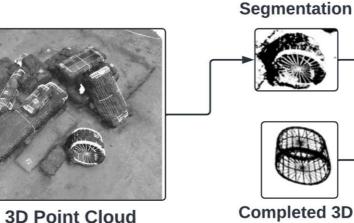
Fig 2: A general grasp detection system involves target object localization, object pose estimation, and grasp estimation.

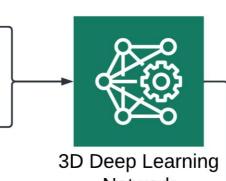


Input Point Clouds End-to-end Prediction Best Grasp Execution

Fig 4: Proposed Pipeline for Training. I will used synthetic training dataset from YCB [2] object set. Given raw still image data from a stereo camera, using PointNet++ based Network can directly predict the poses, categories and scores (qualities) of all the grasps in a fast way [3].



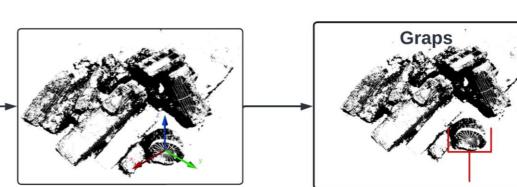




**3D Point Cloud** 

Shape

Network



**Direct Regress 6D Poses** 

Fig 5: Functional flow-chart of 3D template-based 6D object pose estimation and grasp methods based on the complete shape.

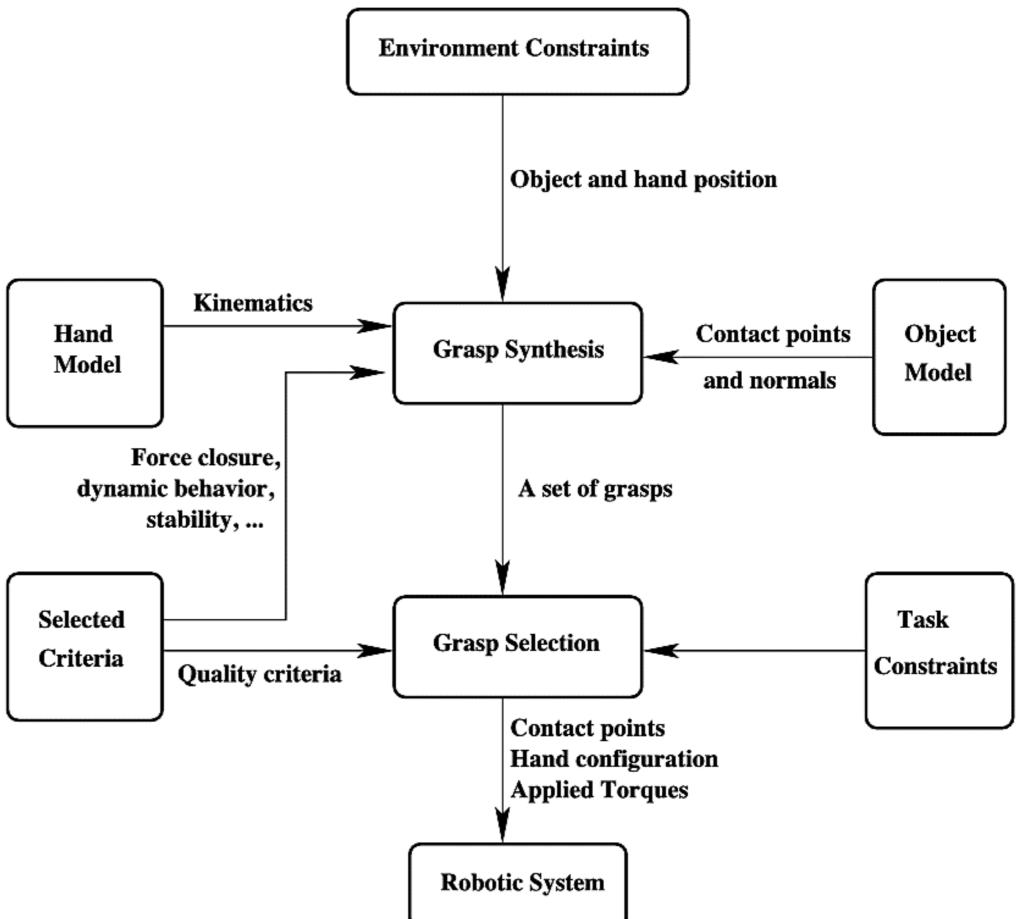


Fig 3: Current state of art underwater autonomous and dexterous operation robot, environment perception, underwater vehicle-manipulator system modeling and coordinated control, target uninjured grasp [1].

## REFERENCES

- 1. Huang, Hai, et al. "A review on underwater autonomous environmental perception and target grasp, the challenge of robotic organism capture." Ocean Engineering 195 (2020): 106644.
- 2. Calli, Berk, et al. "Benchmarking in manipulation research: Using the Yale-CMU-Berkeley object and model set." IEEE Robotics & Automation Magazine 22.3 (2015): 36-52.
- 3. Ni, Peiyuan, et al. "Pointnet++ grasping: Learning an end-to-end spatial grasp generation algorithm from sparse point clouds." 2020 IEEE International Conference on Robotics and Automation (ICRA). IEEE, 2020.





Fig 6: Strategy of grasp synthesis using analytical approaches.





UNIVERSITÉ DE 

