

Learning Visual Shape Control of Novel 3D Deformable Objects from Partial-View Point Clouds

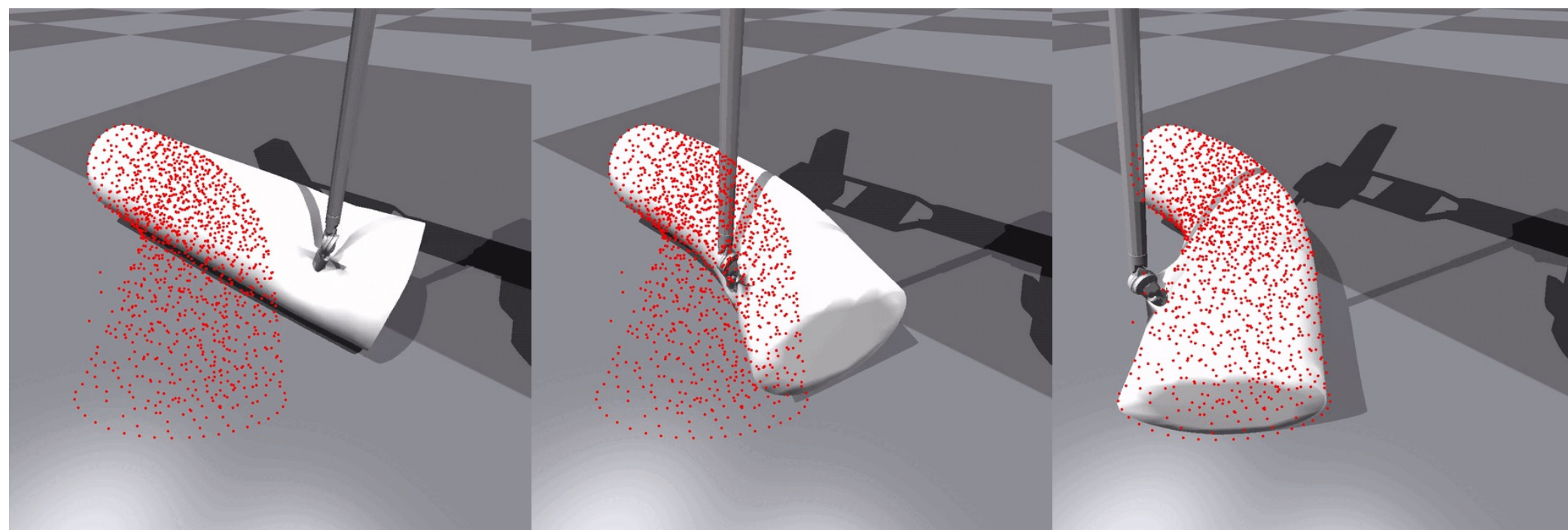
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 Project website: <https://sites.google.com/view/deformernet/home>

Problem

- How can a robot control the shapes of 3D deformable objects?

Motivation

- Deriving a low-dimensional but expressive state representation is difficult due to infinite DOFs.
- Prior works:
 - Points on the object surface
 - Simple feature descriptor of object point cloud
- These methods over-simplify the *complex dynamics* of 3D deformable objects.
- Why not learn it via deep neural networks?



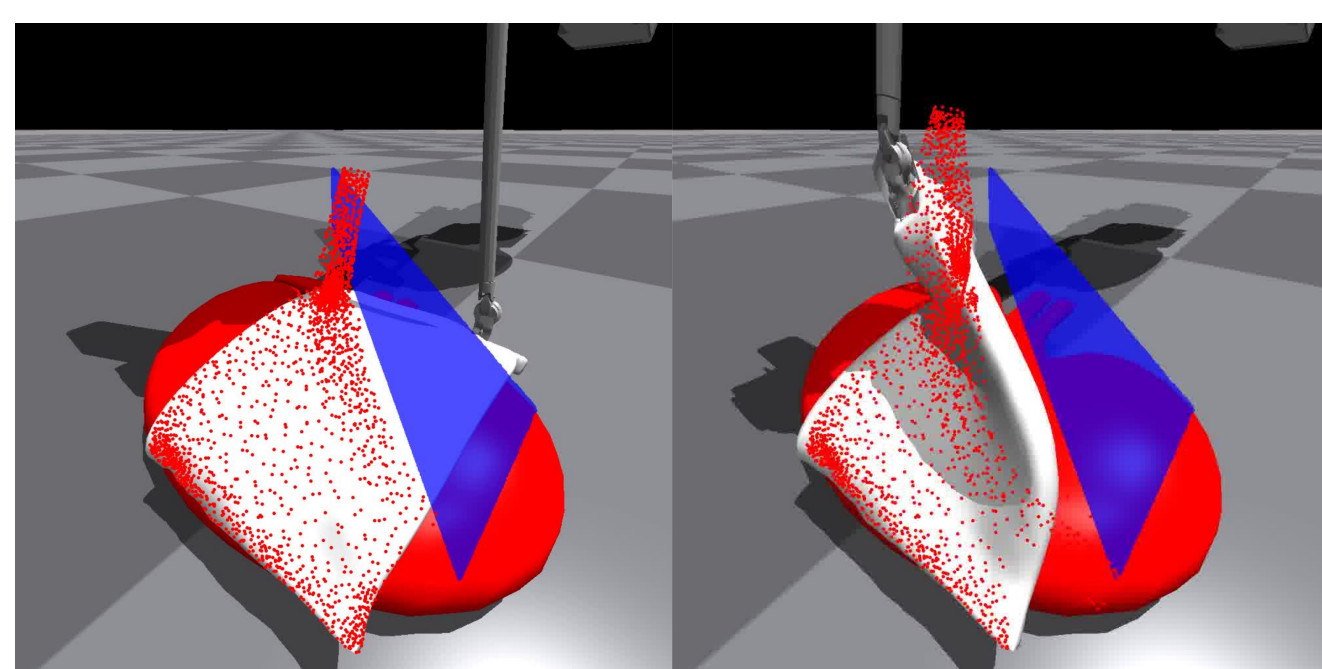
Broader Impact

Critical to many robotic manipulation tasks:

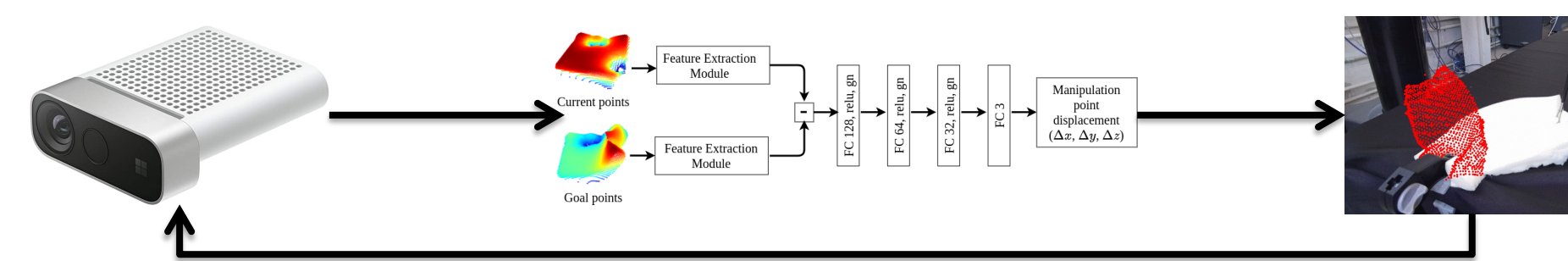
- Surgical assistive robots: tissue and organs
- Home-assistance robots: sponges, clothes, food, etc.
- Warehouse robots: containers, boxes, etc.
- Safe automated productions lines for deformable products.

Surgical Application

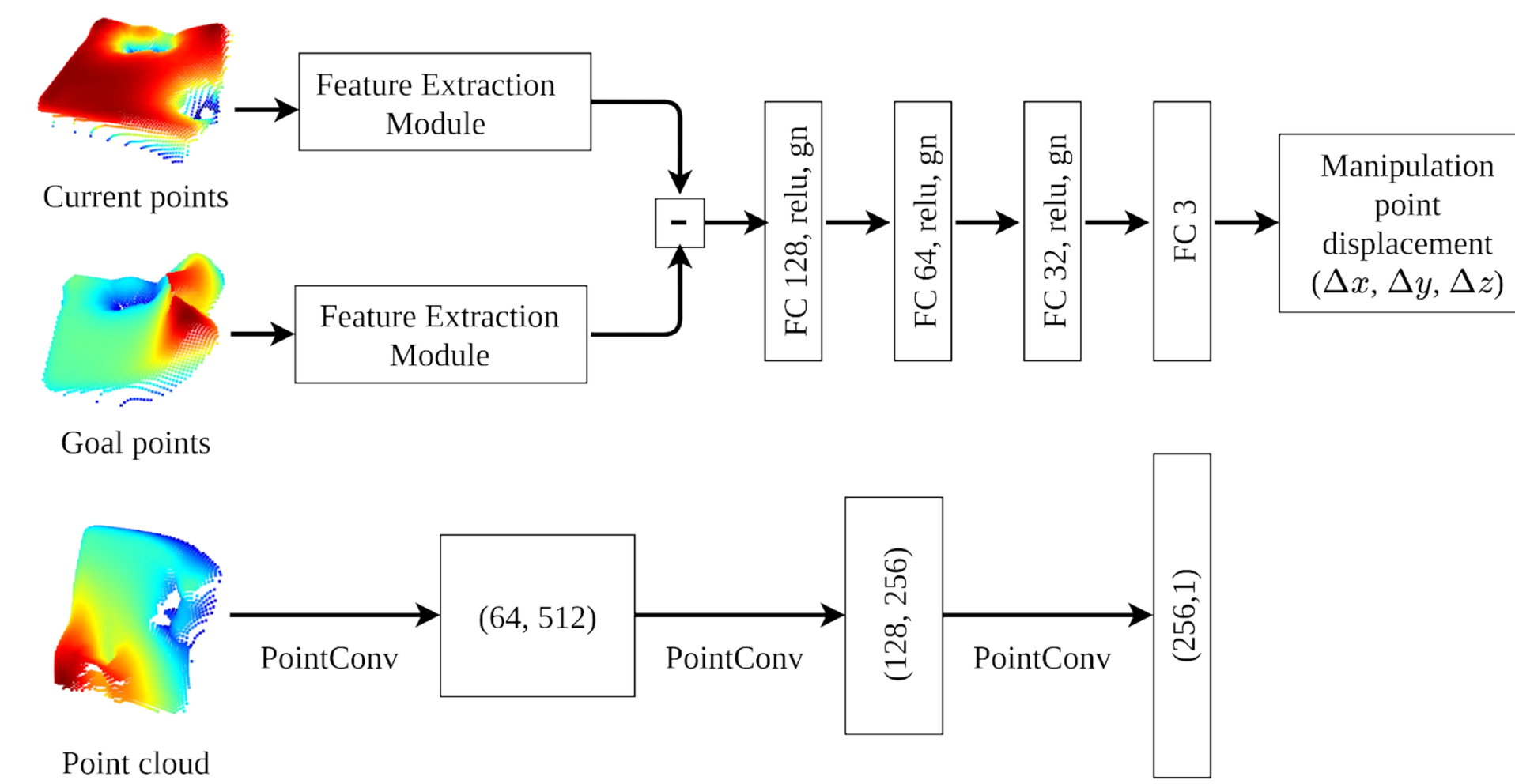
Surgical retraction task: grasping a tissue layer and lifting it up to expose the underlying area of a kidney.



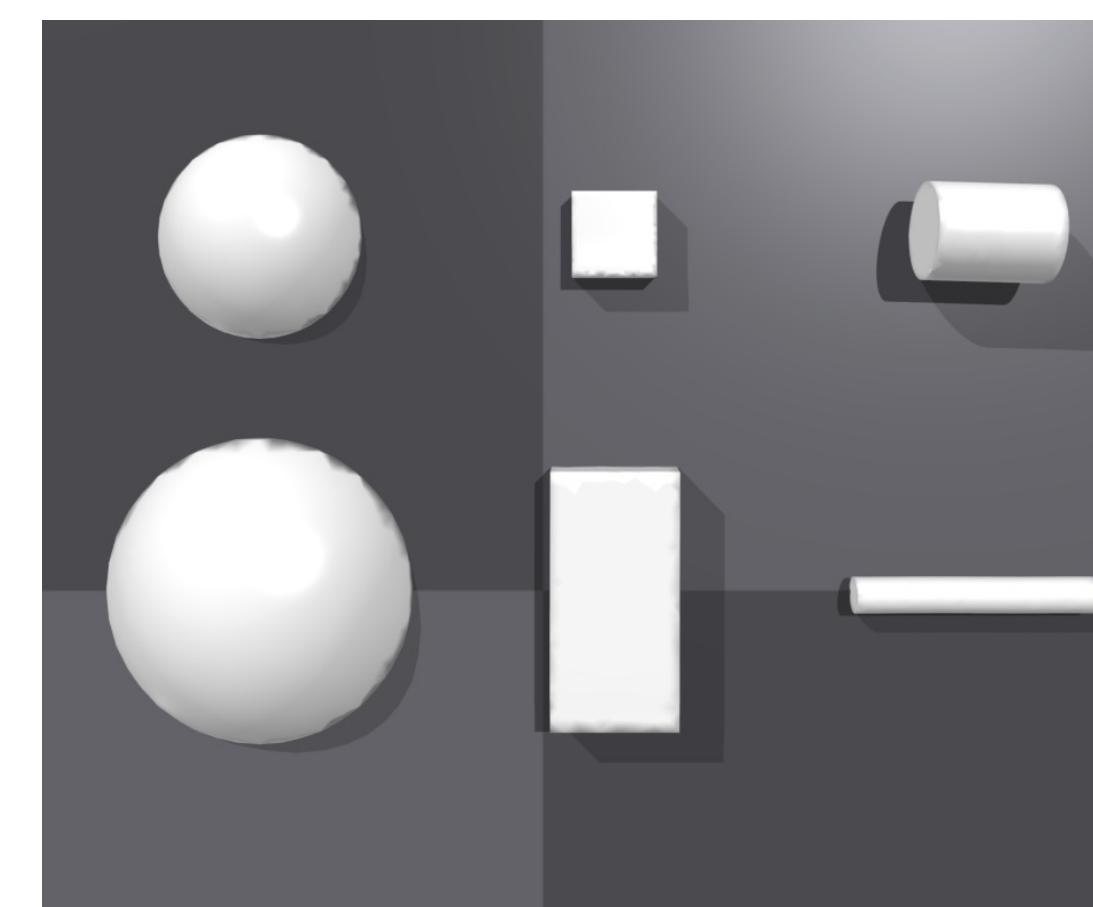
Shape Servoing with *DeformerNet*



- Obtain point cloud → Pass to DeformerNet to compute action → control robot → repeat until convergence

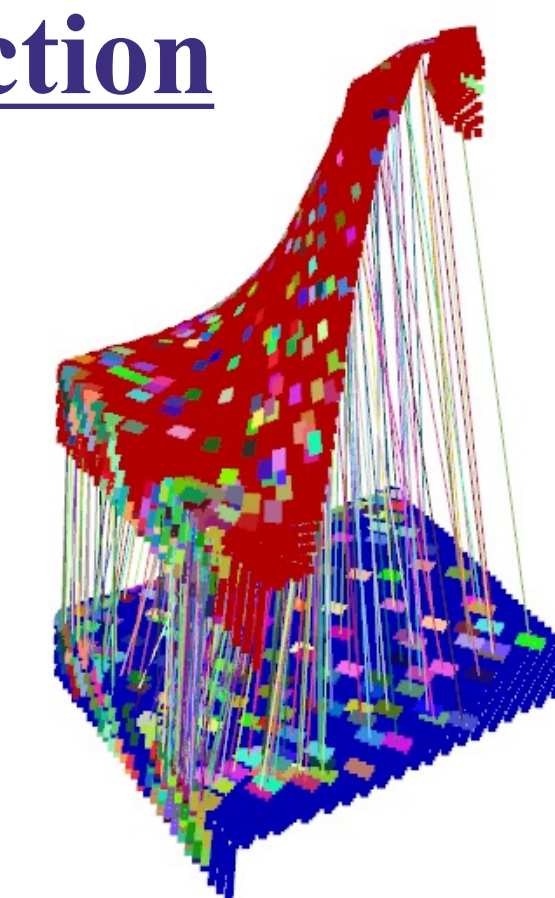


- Feature extraction: generate a 256-dim feature vector from partial-view point cloud using PointConv
- Deformation control: compute end effector displacement given difference of current and target features
- We train on random size and stiffness instances of these shapes



Manipulation Point Prediction

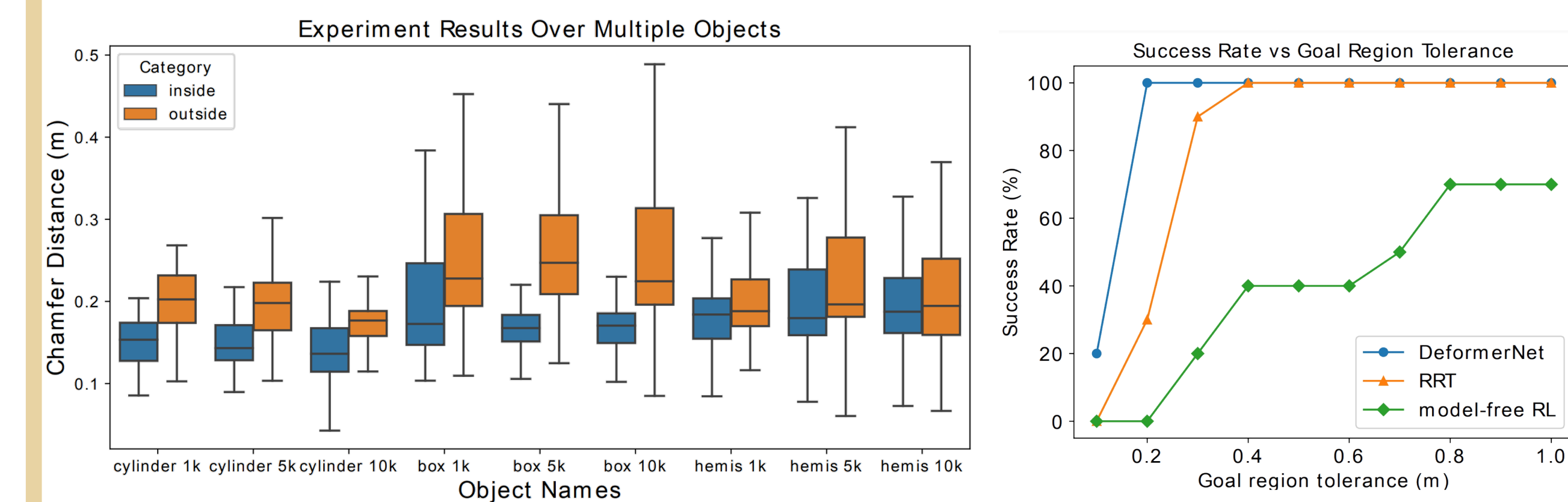
- Where to grasp when performing shape control?
- Insight: points on the object that move more should generally be close to the grasp point
- Manipulation point = displacement-weighted average of keypoints matched between initial and target shape.



Experiments

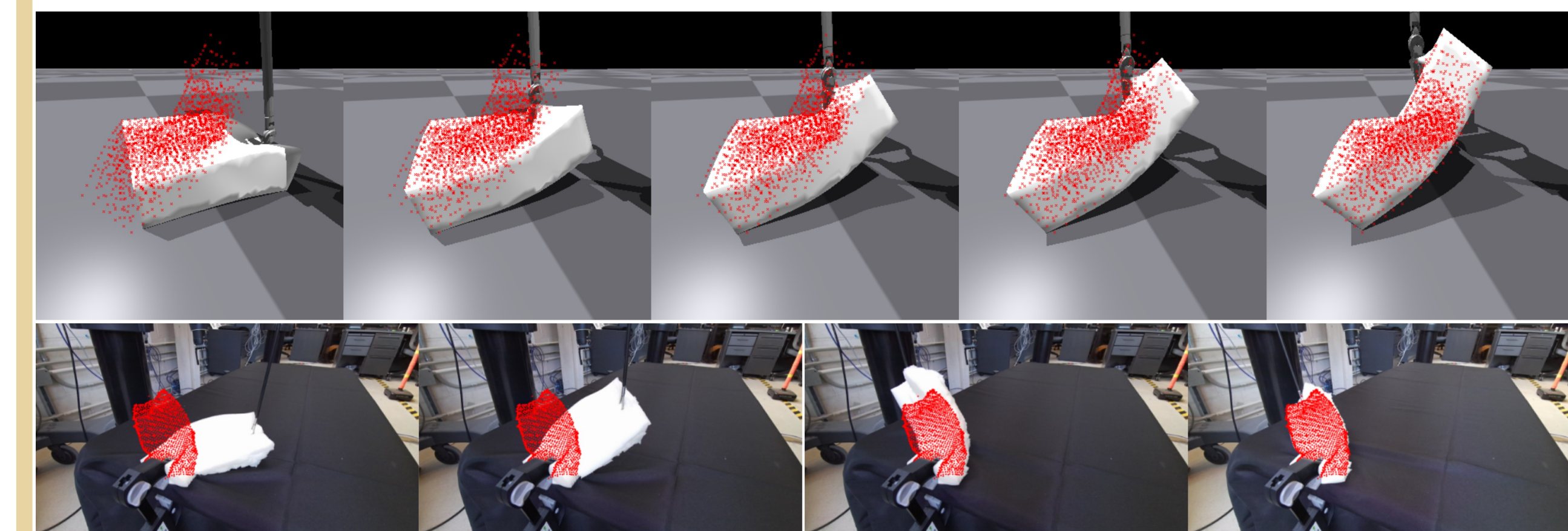
Quantitative Results

- In simulation, use dVRK surgical robot to manipulate a variety of object geometries and stiffnesses.
- Evaluate performance in 2 categories: objects inside and outside the training distributions

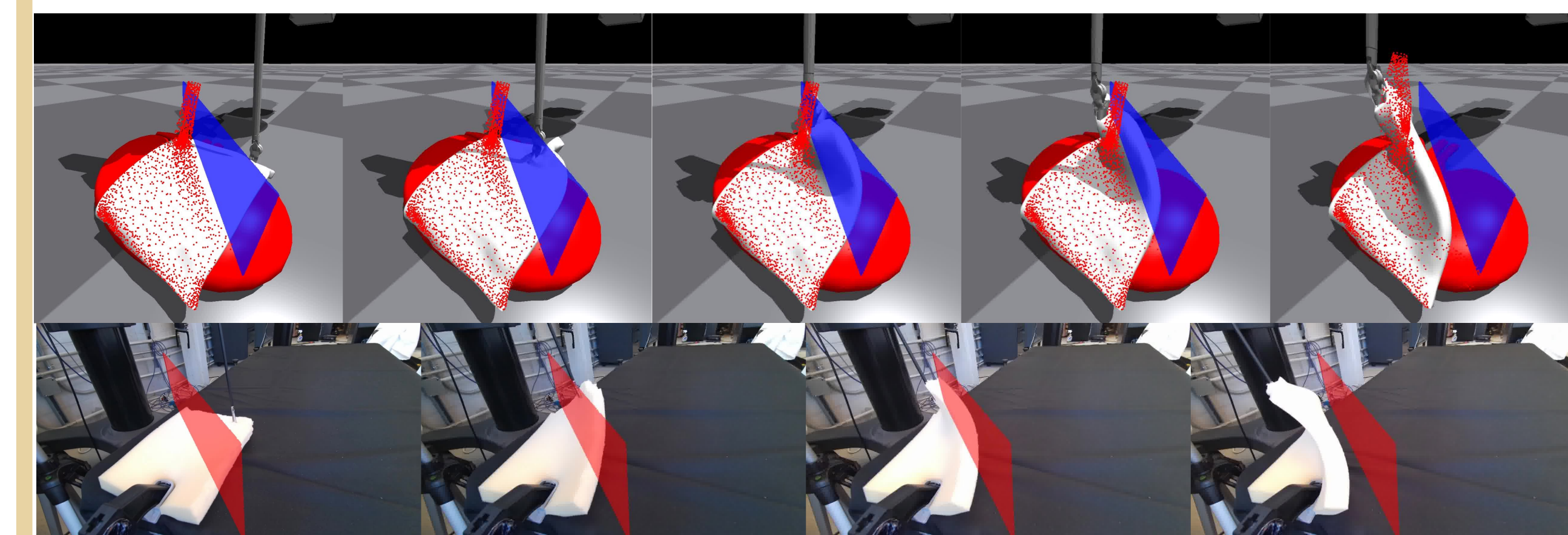


- On real robot, evaluate with a foam box object.

Qualitative Results



Goal-oriented shape servoing



Surgical retraction task