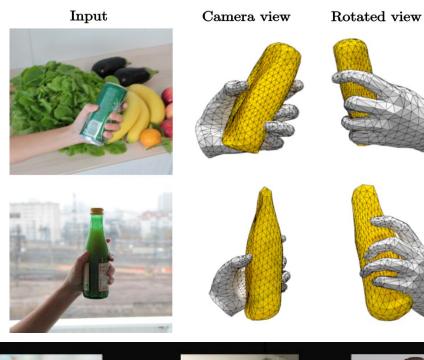


Learning joint reconstructions of hands and manipulated objects







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Michael J. Black



Ivan

Laptev



Cordelia Schmid

Motivation





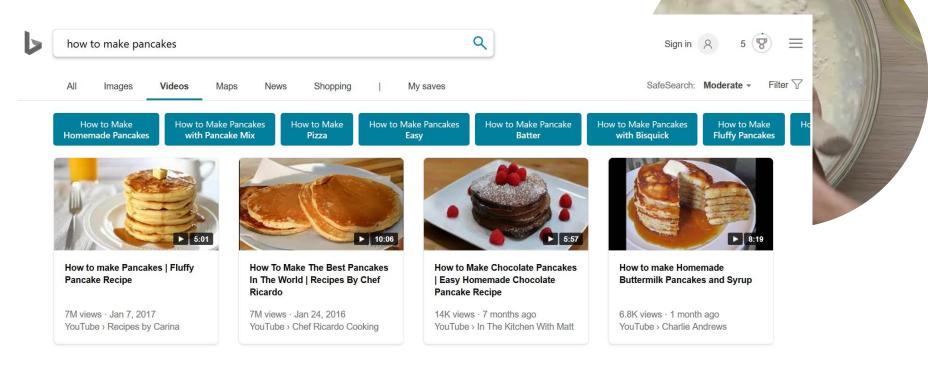
Motivation

- Learn human rules of object manipulation
- A first step : hand-object reconstruction from a single rgb frame



Motivation

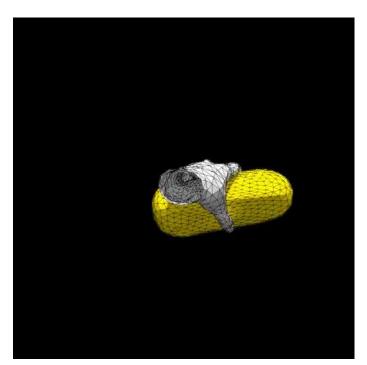
- Learn human rules of object manipulation
- A first step : hand-object reconstruction from a single rgb frame



Our approach

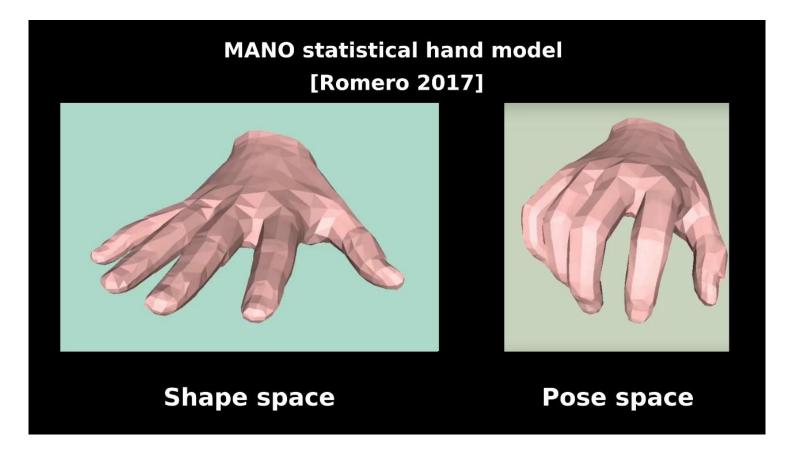
 Reconstruct hand and object meshes in an end-to-end learning framework





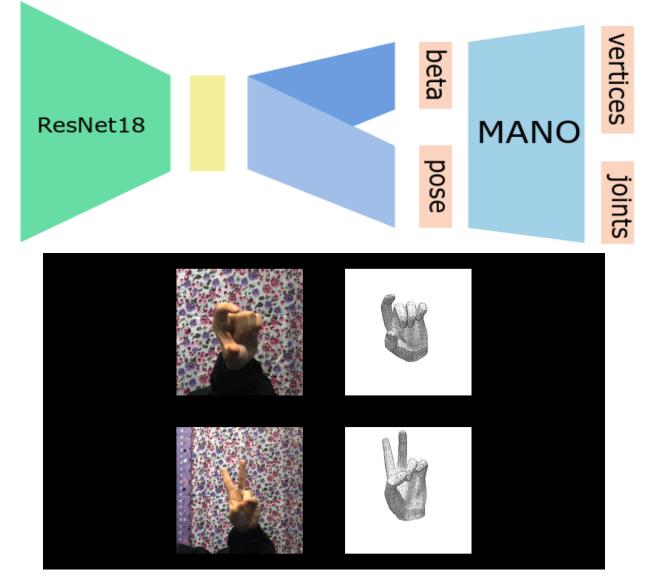
Hand reconstruction

• Integrate the <u>MANO</u> parametric hand model as a differentiable layer



Hand reconstruction

• Integrate the MANO parametric hand model as a differentiable layer

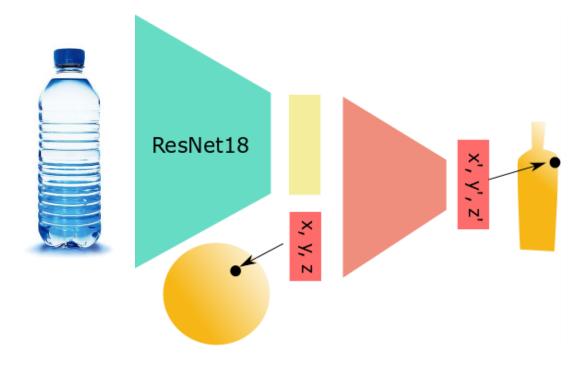


Object reconstruction



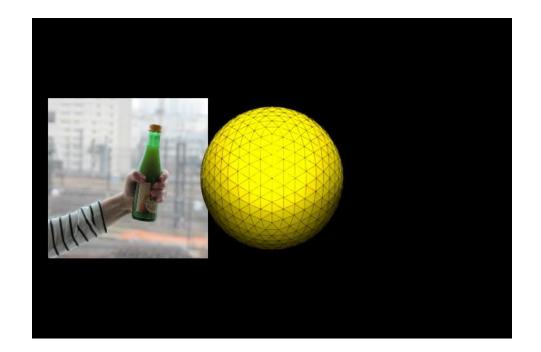
- Deform a sphere into the target shape
- Baseline from <u>AtlasNet</u>



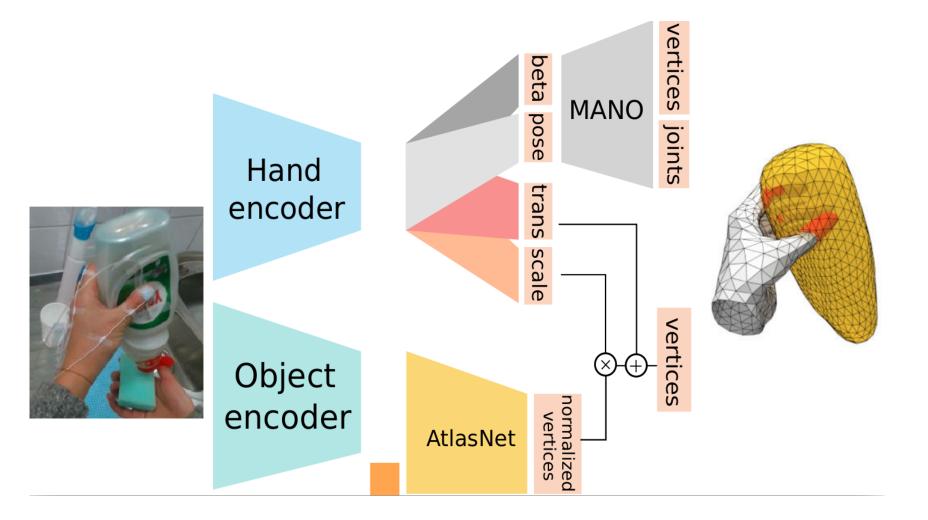


Object reconstruction

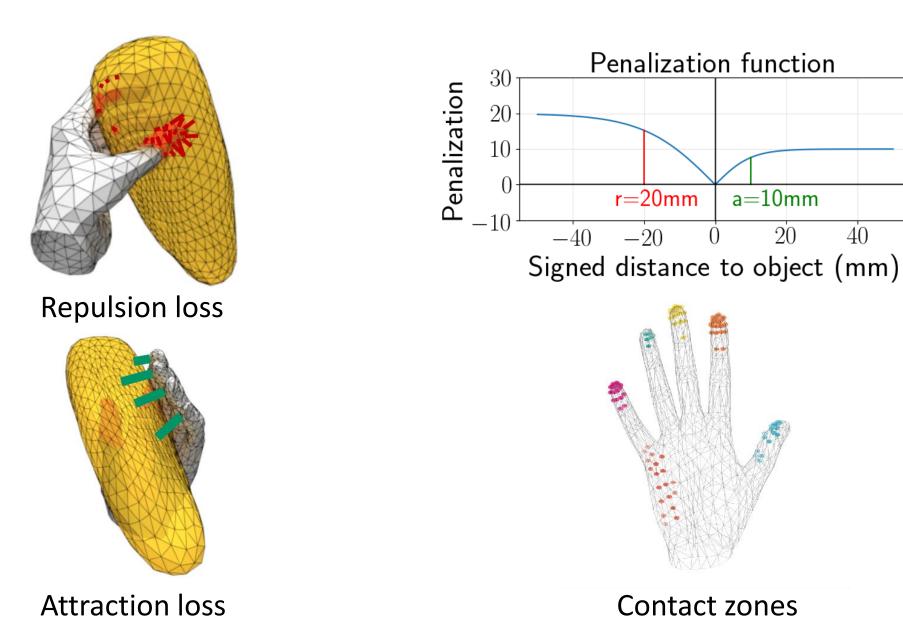
 Predict scale and translation relative to hand



Full architecture

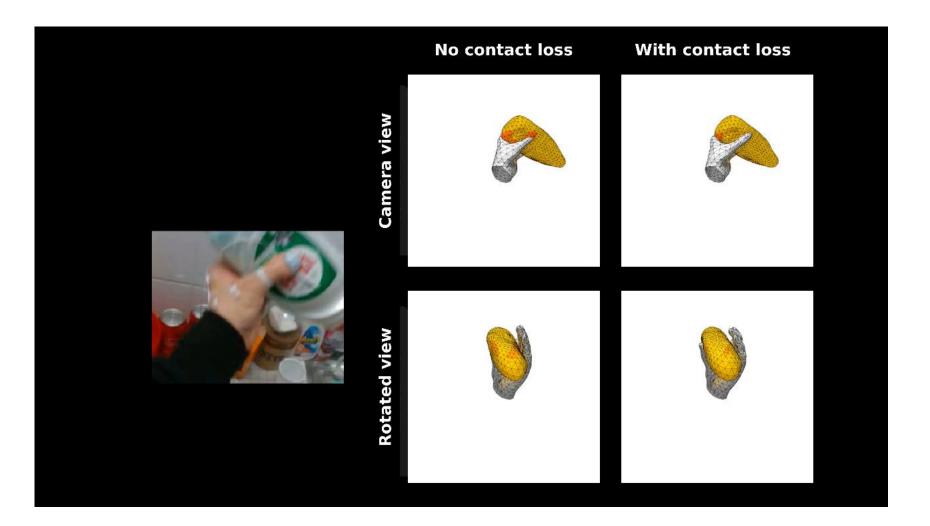


Interaction terms



Qualitative results

First Hand Action Benchmark dataset



Object model collection

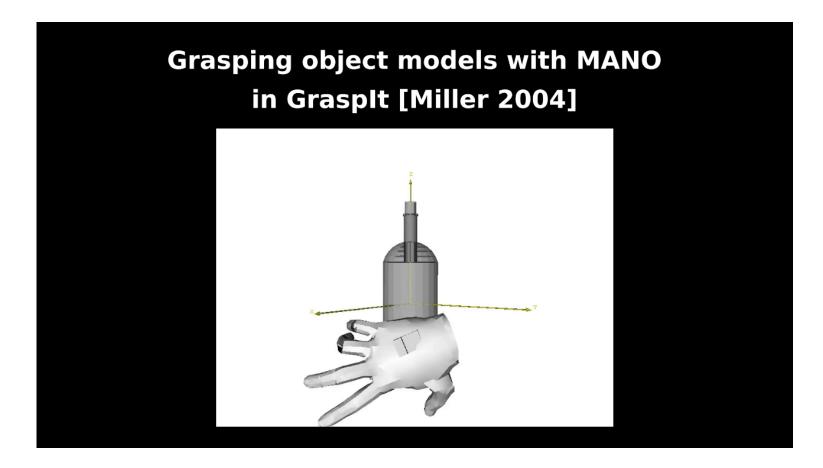
Selected "Graspable" objects from <a>ShapeNet

- 8 object categories (bowls, bottles, ...)
- 2.7K object instances



Robotic grasps

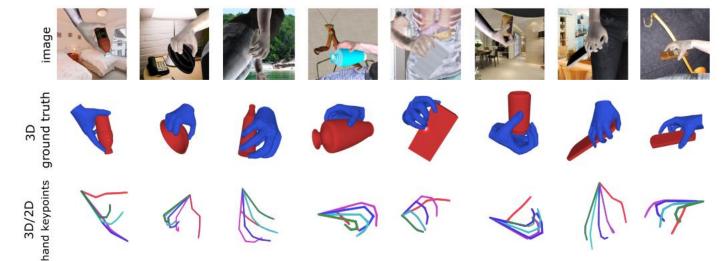
Transformed MANO to a fixed articulated model Use GraspIt software to automatically generate grasp



Rendering

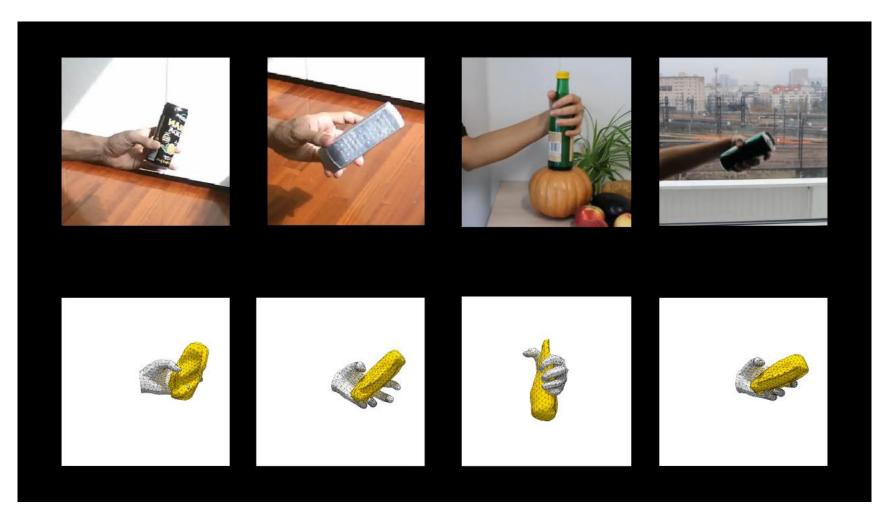
Random object textures from ShapeNet and image backgrounds Random lighting, body pose, global orientation

Samples from the generated ObMan dataset



Qualitative results

<u>CORe50</u> dataset and custom sequences



Failure cases



Model robustness

Background variation



Object textures



Unseen categories



Future work...

