

## The System

Generates a direct mapping from a scene to a representation of grasp affordances.

Anti-podal grasps are encoded as a likelihood of success (Grasp Score) and Grasp Angle for each input pixel.

Transformed into 6-DoF grasp poses using depth and surface normals.

Preliminary results on Cornell Grasping Dataset shown below (grasps extracted by locating local maxima in the Grasp Score).

## Advantages

No sampling of grasp candidates – grasps are represented as a continuous space.

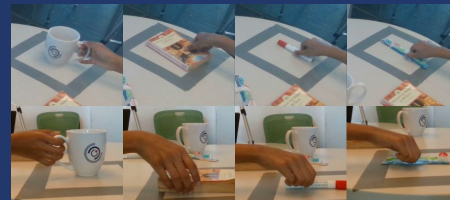
Better represents a robot's environment in terms of the distribution of grasp affordances compared to sampling/discretising the grasping space or generating a single grasp pose per scene.

Future work: integration to existing robotics techniques such as visual servoing or active perception – no more blind grasping.

## Extending to multi-fingered grasping

Future work: Learn to generate grasps by encoding finger-wise placement models – not possible with commonly used rectangle representation.

We have collected a video dataset of human grasping for this purpose.



## One-to-one generative mapping from image-space to grasping-space.

