

Learning about Objects by Experiment



— Paul Fitzpatrick —

MIT Artificial Intelligence Laboratory

machine perception

algorithms

+

training data

=

robust
perception

self-training perceptual system

algorithms

+

opportunities

=

training data

+

more
algorithms

=

robust
perception

making opportunities

In robotics, vision is often used to guide manipulation

But manipulation can also guide vision

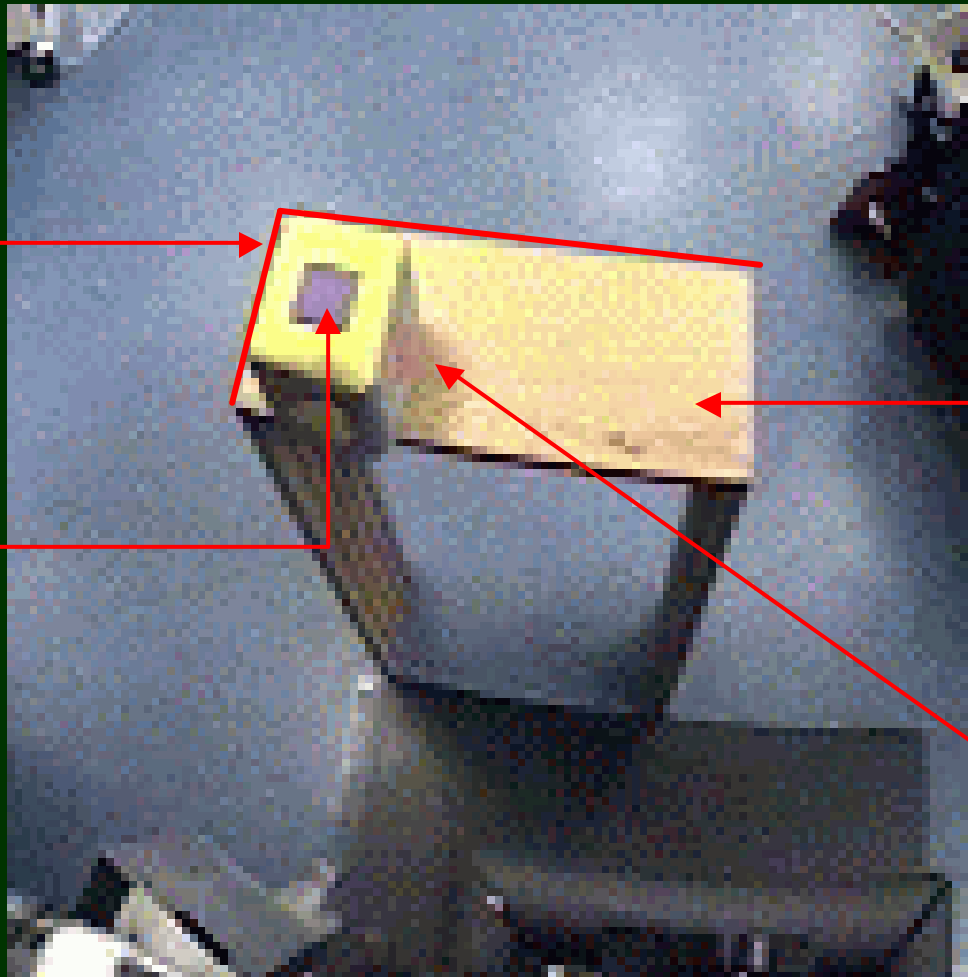
Important for...

- Experimentation progressing when perception is ambiguous
- Correction recovering when perception is misleading
- Development bootstrapping when perception is primitive

a simple scene?

Edges of table
and cube
overlap

Cube has
misleading
surface pattern



Color of cube and
table are poorly
separated

Maybe some cruel
grad-student
faked the cube
with paper,
or glued it
to the table

active segmentation

- Object boundaries are not always easy to detect visually

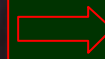
active segmentation

- Object boundaries are not always easy to detect visually
- Solution: Cog sweeps through ambiguous area



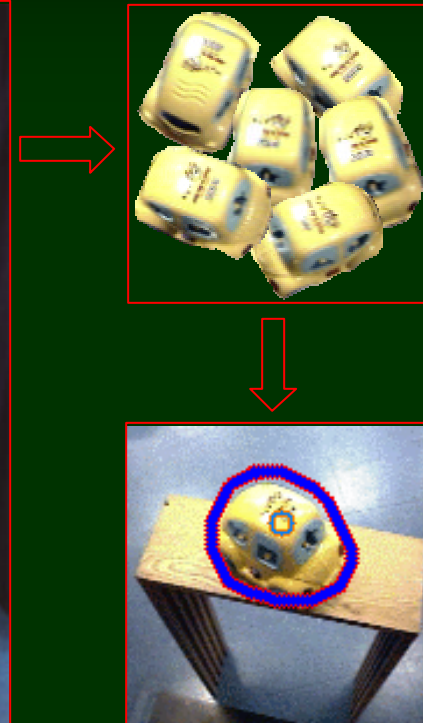
active segmentation

- Object boundaries are not always easy to detect visually
- Solution: Cog sweeps through ambiguous area
- Resulting object motion helps segmentation



active segmentation

- Object boundaries are not always easy to detect visually
- Solution: Cog sweeps through ambiguous area
- Resulting object motion helps segmentation
- Robot can learn to recognize and segment object without further contact



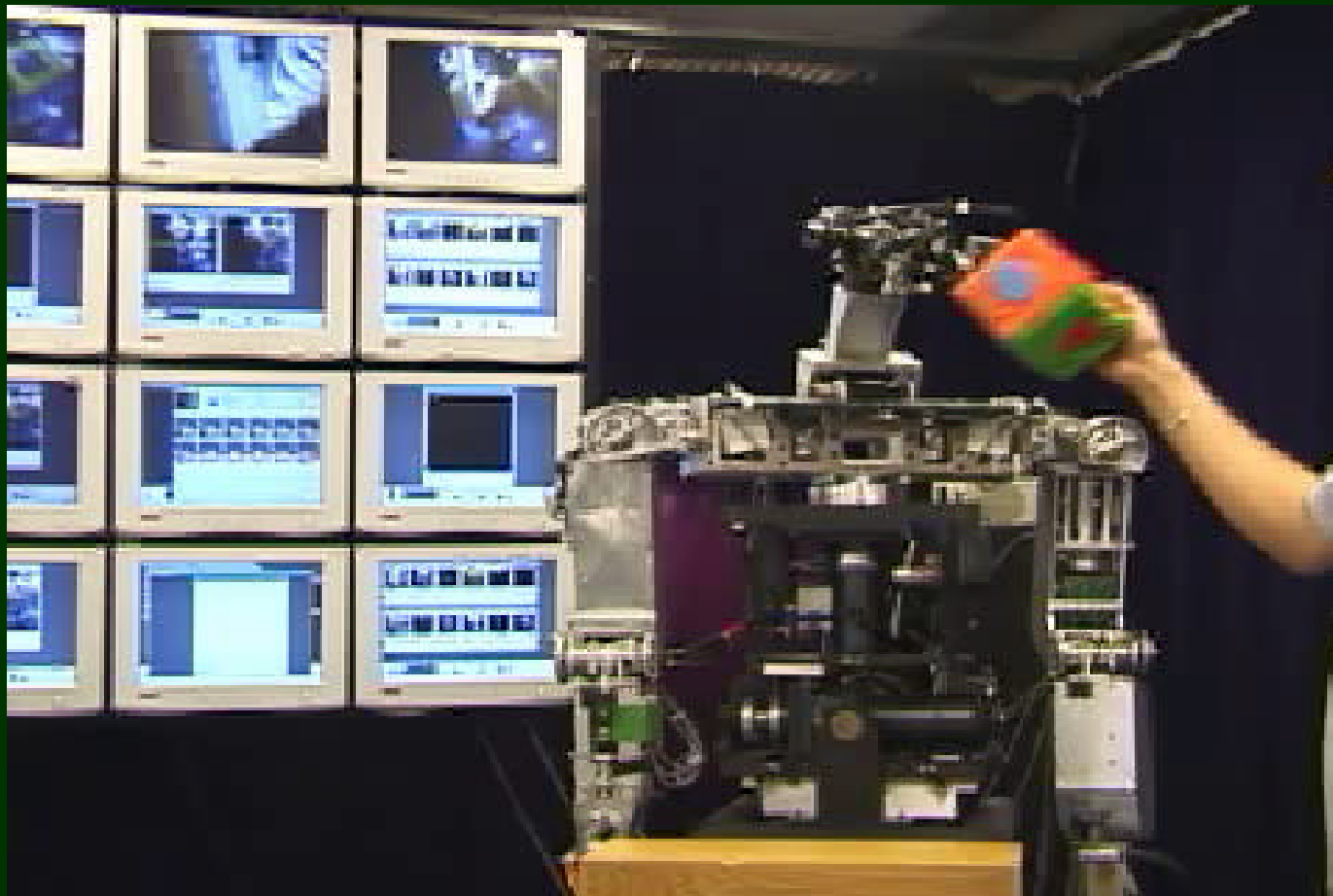
active segmentation



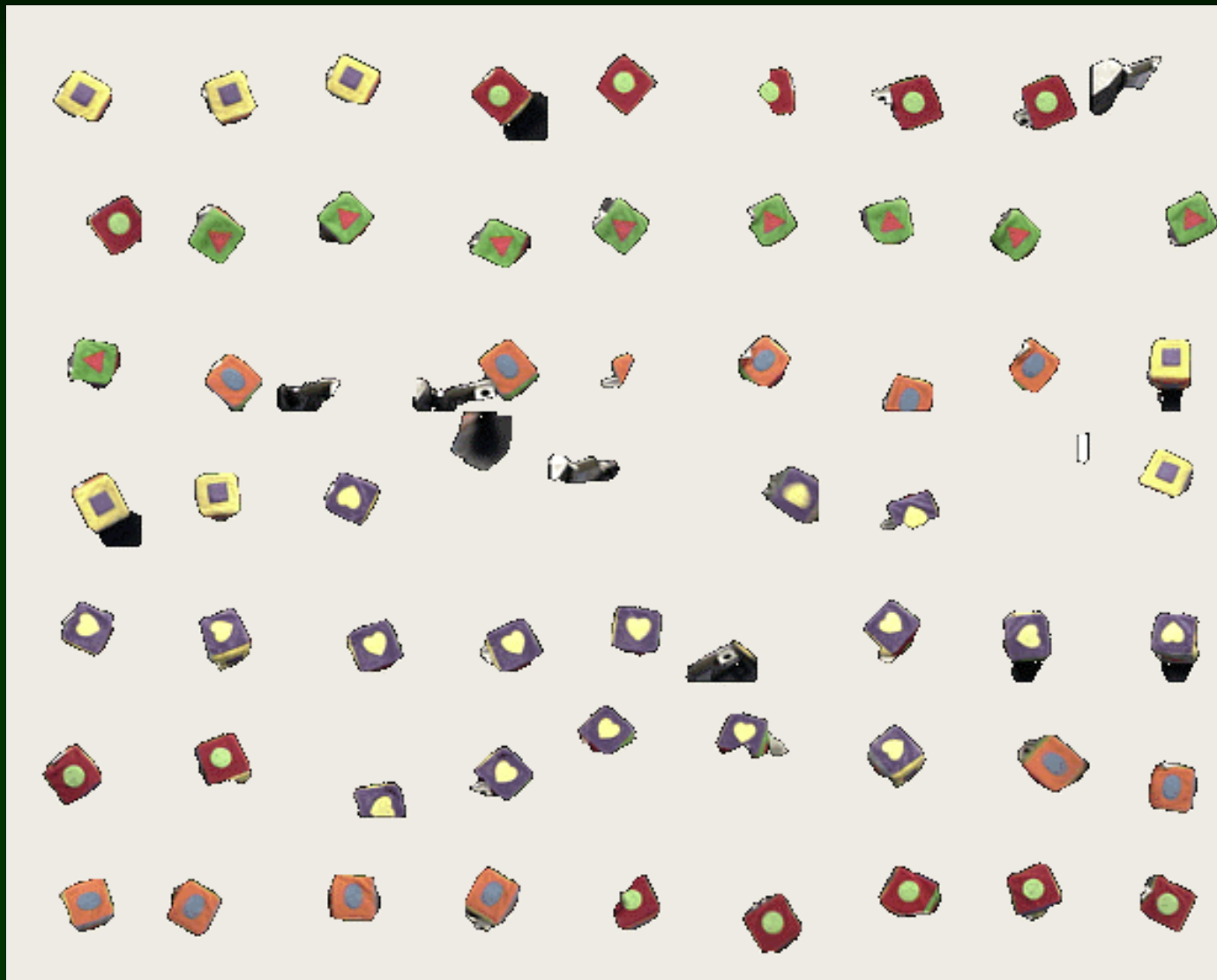
active segmentation



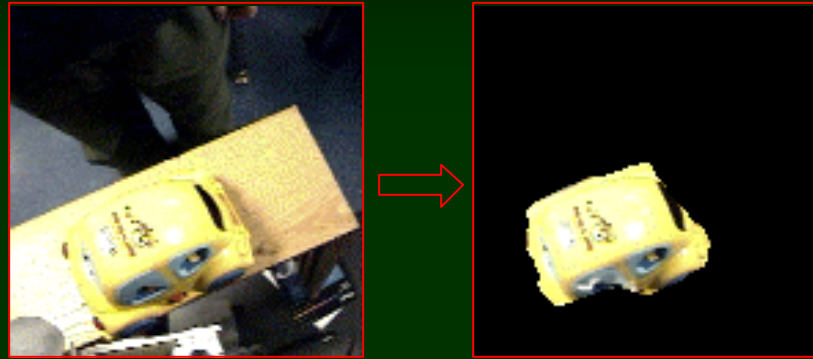
integrated behavior



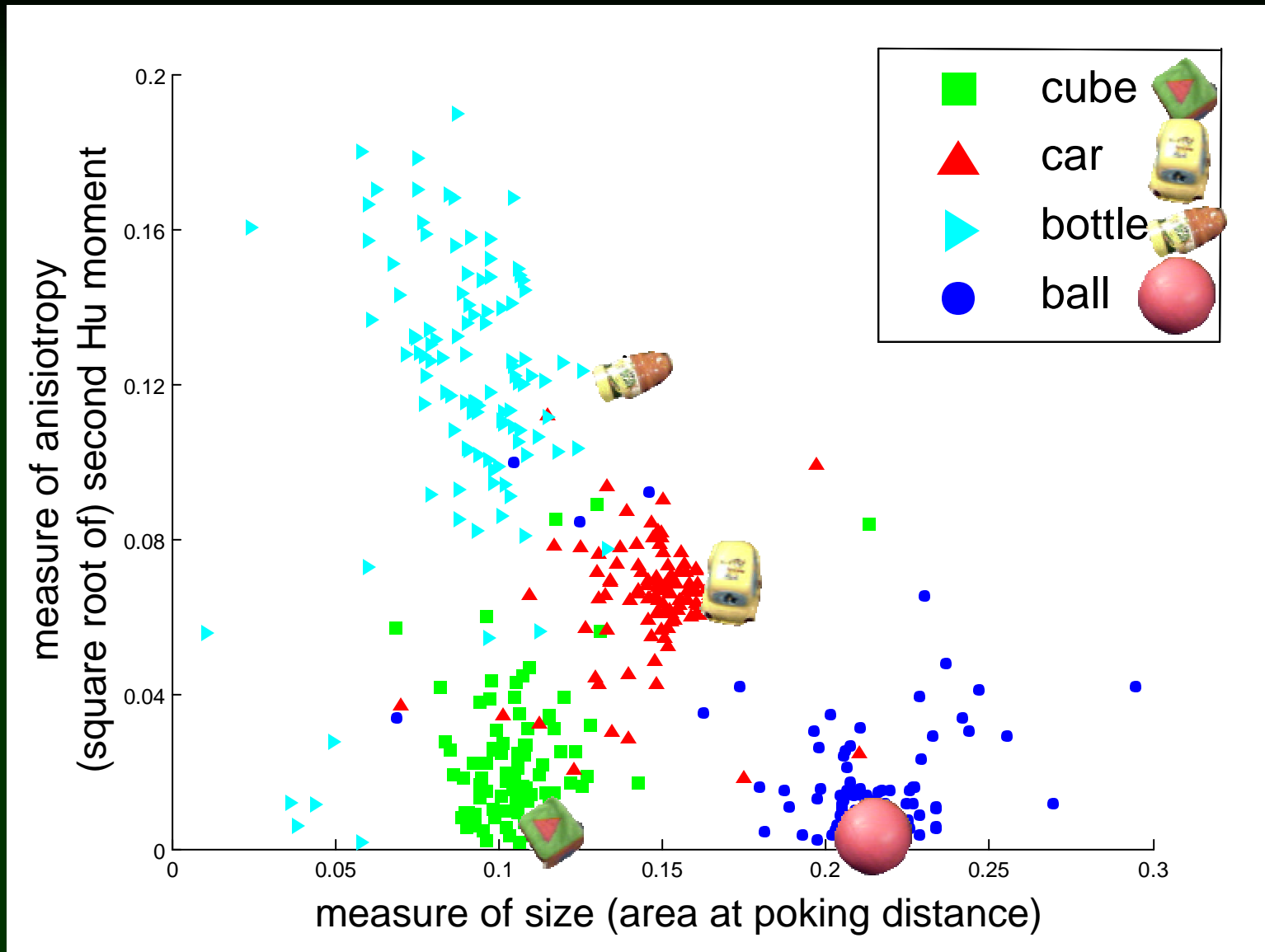
segmentation examples



segmentation examples



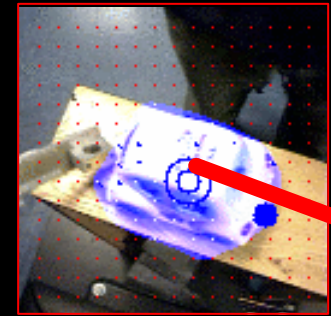
boundary fidelity



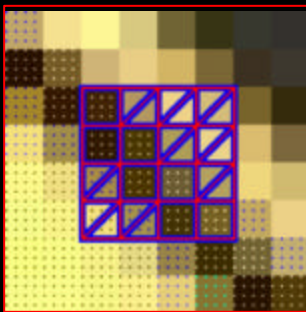
poking

affordance exploitation
(rolling)

object segmentation



edge catalog



object detection
(recognition, localization,
contact-free segmentation)



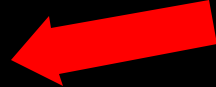
manipulator detection
(robot, human)



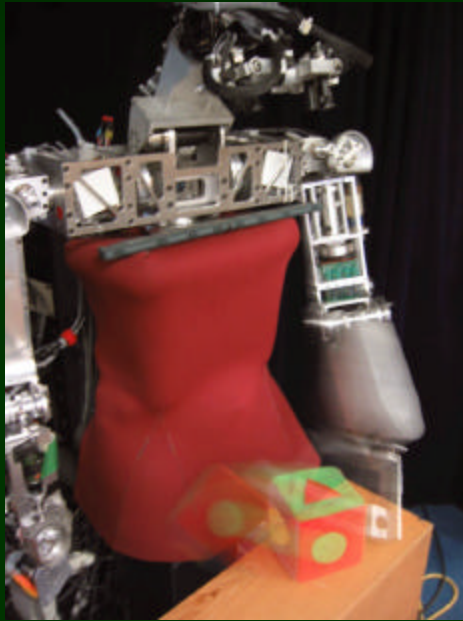
poking



object segmentation



opportunities for segmentation



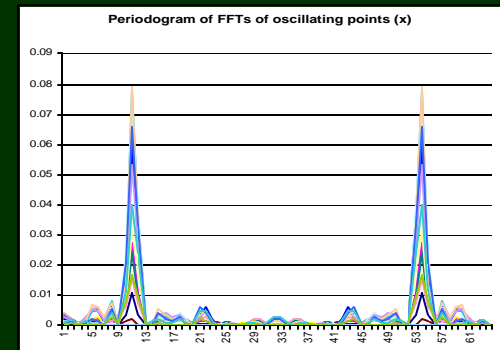
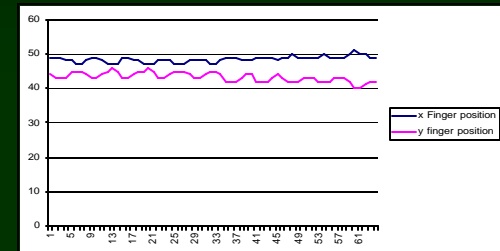
robot, poking

(Paul Fitzpatrick,
Giorgio Metta)



wearable, illumination

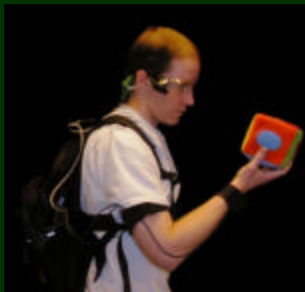
(Charlie Kemp)



camera, periodicity

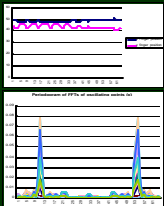
(Artur Arsenio)

segmentation on a wearable



System detects when wearer reaches for an object, requests wearer to hold it up, then illuminates it

segmentation by watching a human



System detects periodic motion – waving, tapping, etc. – and extracts seed points for segmentation

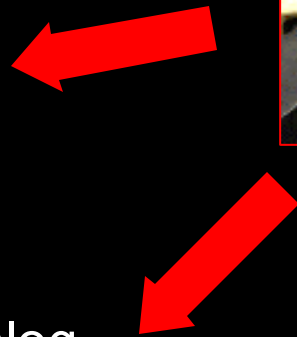
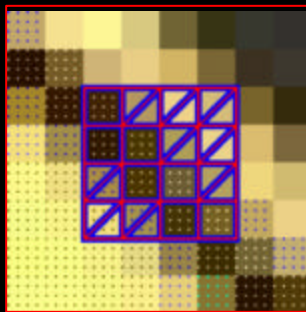
poking



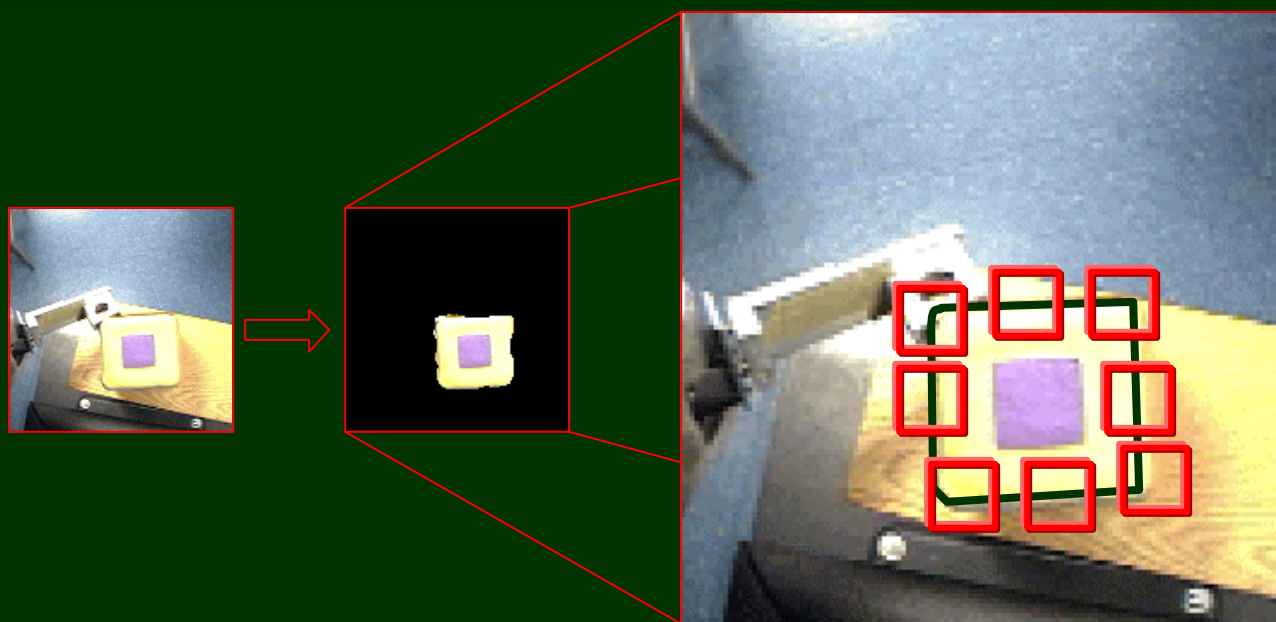
object segmentation



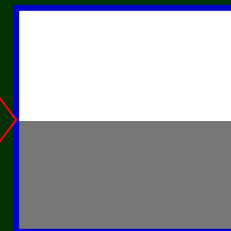
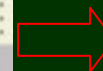
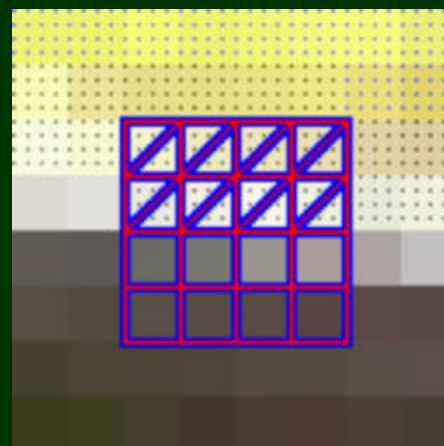
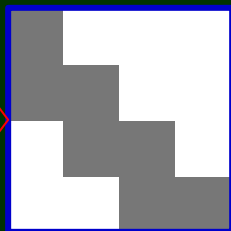
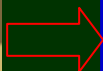
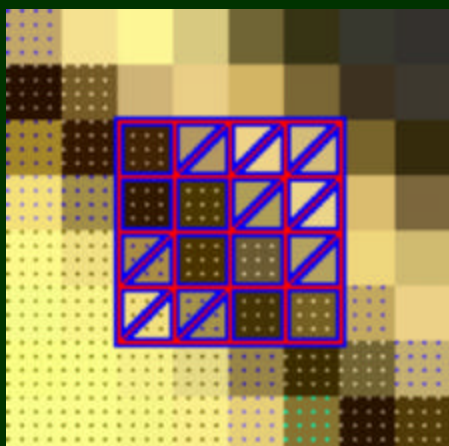
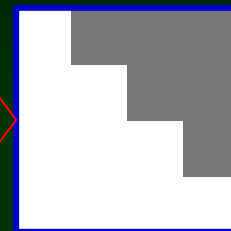
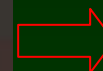
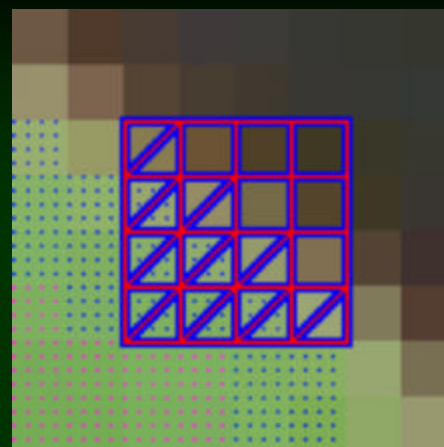
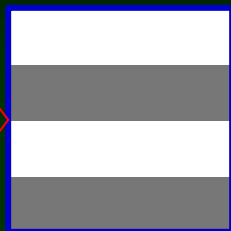
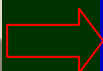
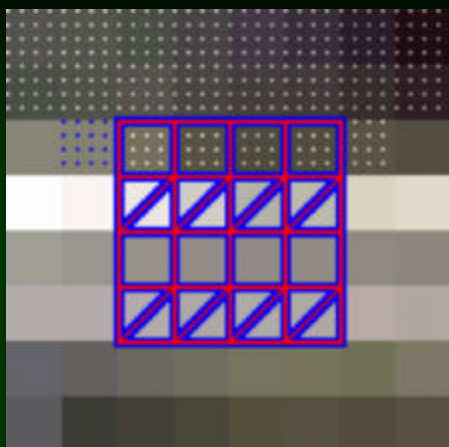
edge catalog



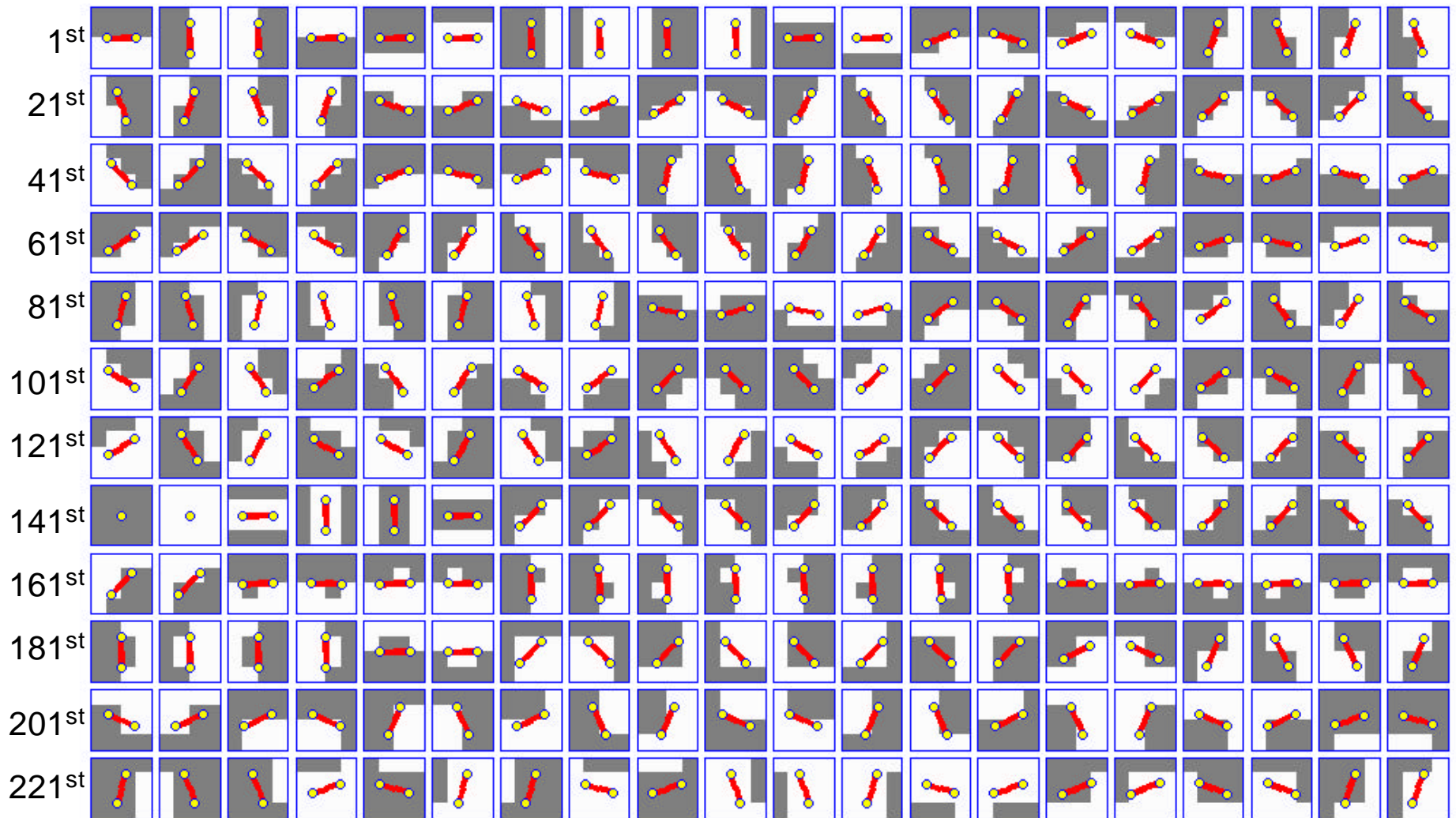
sampling oriented regions



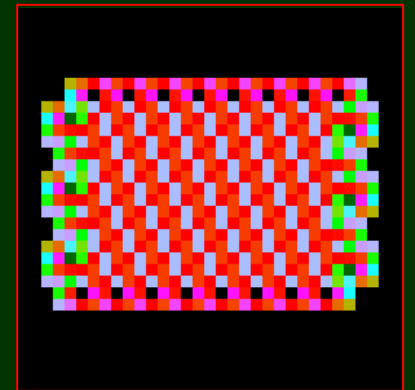
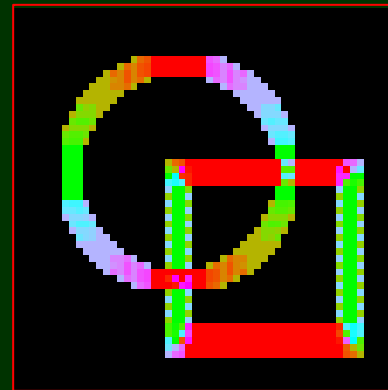
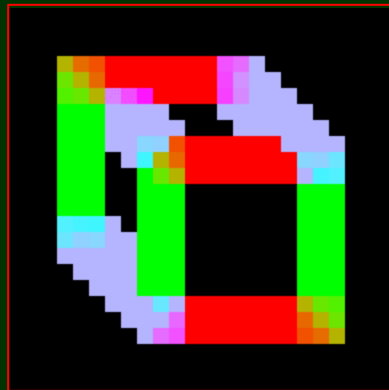
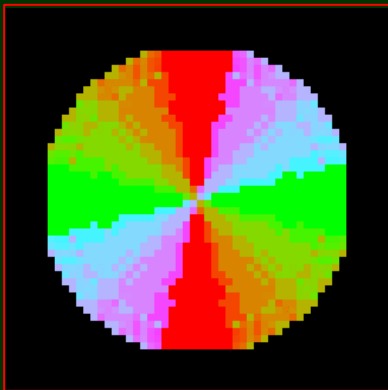
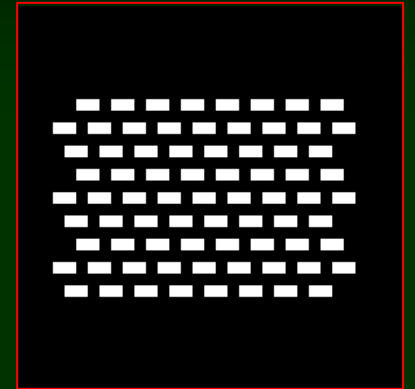
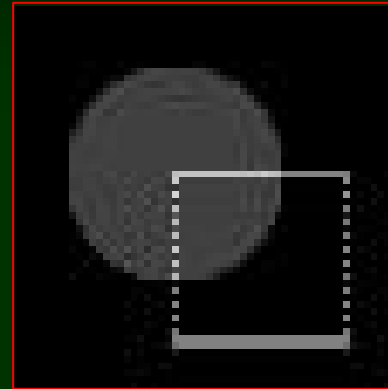
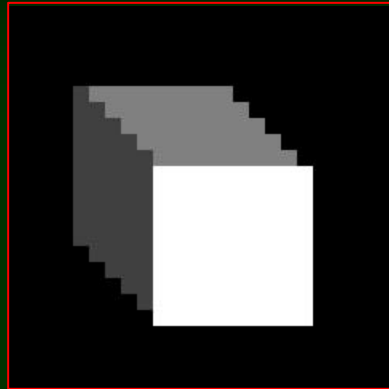
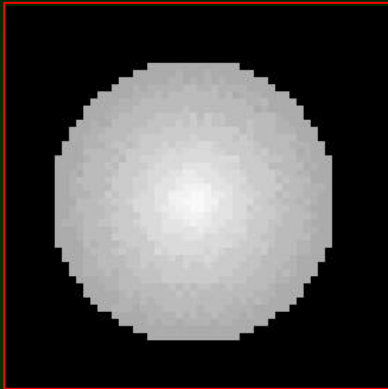
sample samples



most frequent samples



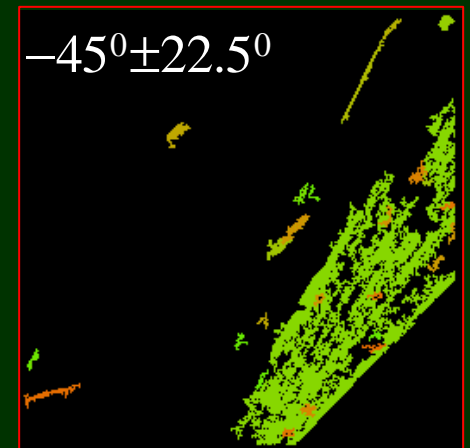
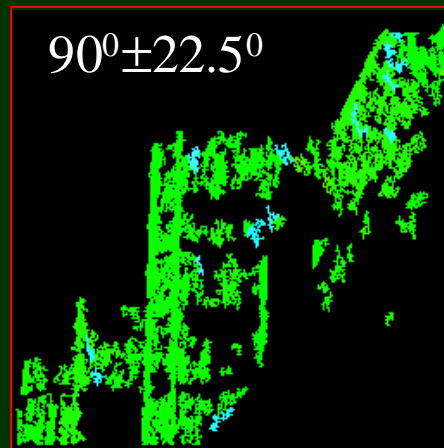
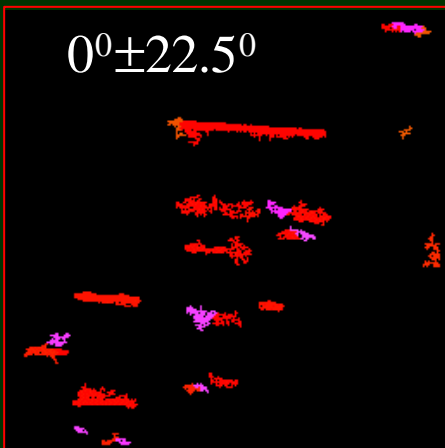
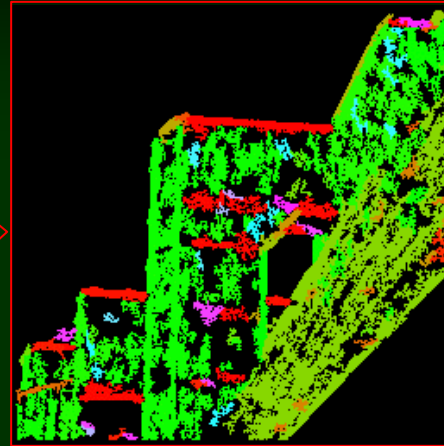
some tests



Red = horizontal

Green = vertical

natural images



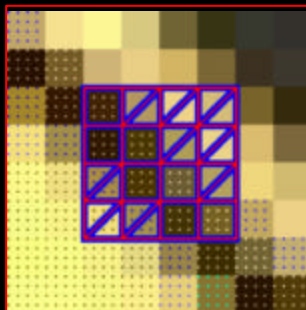
poking



object segmentation



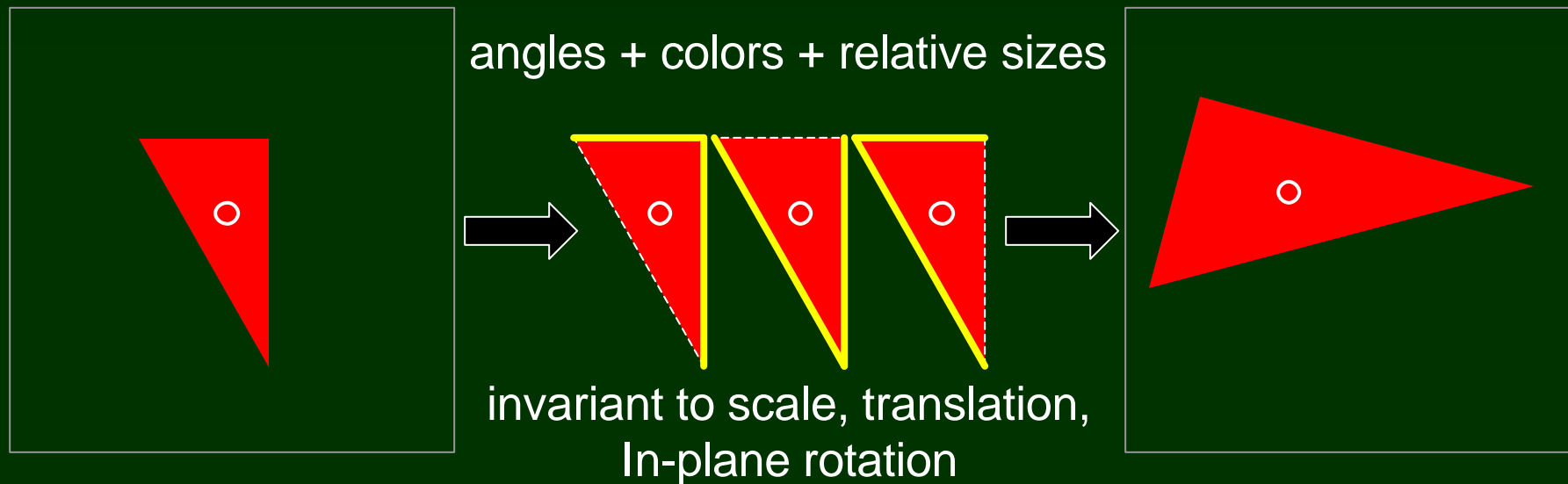
edge catalog



object detection
(recognition, localization,
contact-free segmentation)



geometry+appearance

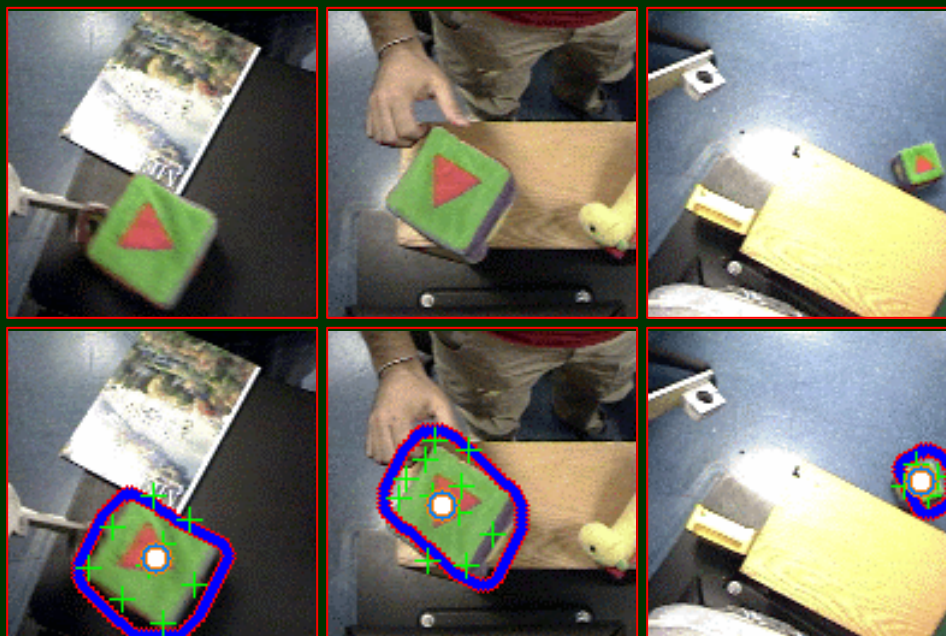


Advantages: more selective; fast

Disadvantages: edges can be occluded; 2D method

Property: no need for offline training

in operation



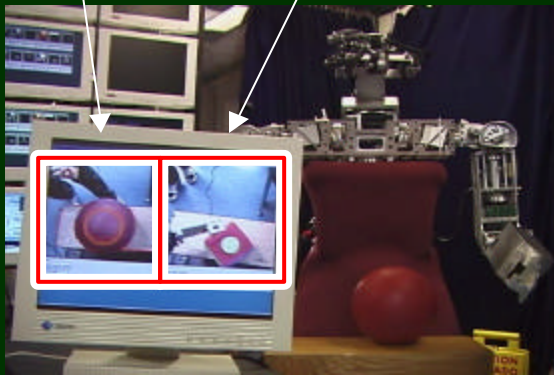
yellow on yellow



open object recognition

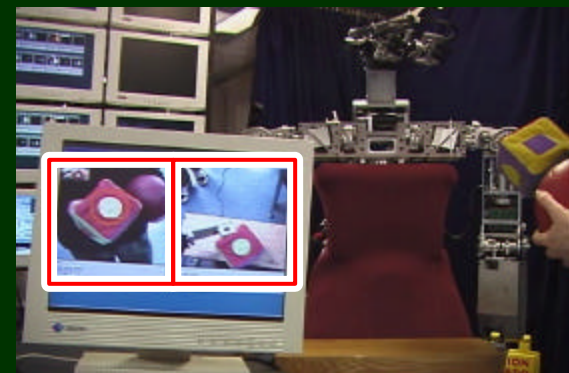
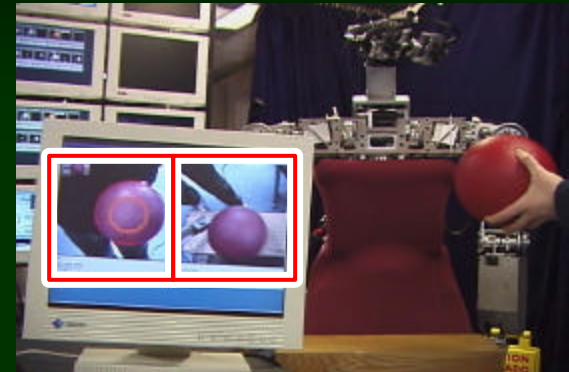
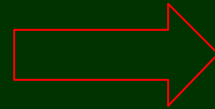
robot's
current
view

recognized
object (as seen
during poking)



sees ball,
"thinks" it is cube

pokes,
segments
ball



correctly differentiates
ball and cube

open object recognition



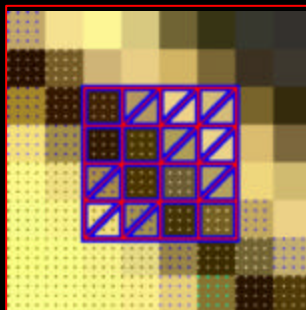
poking



object segmentation



edge catalog



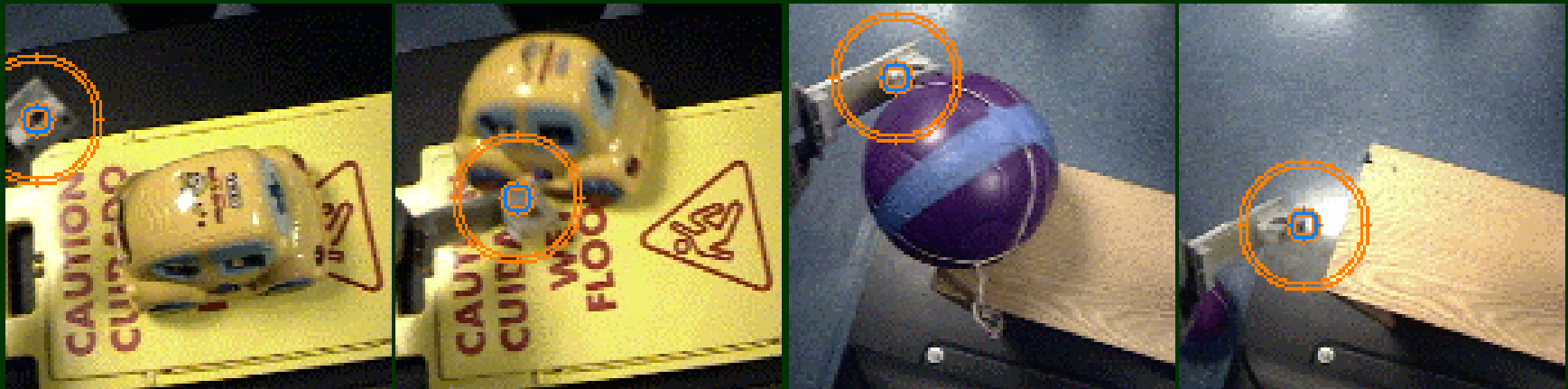
object detection
(recognition, localization,
contact-free segmentation)



manipulator detection
(robot, human)



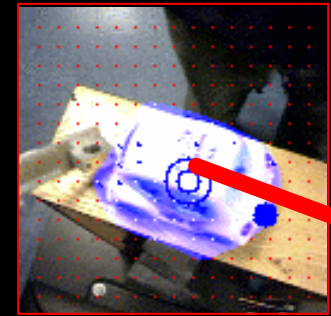
manipulator recognition



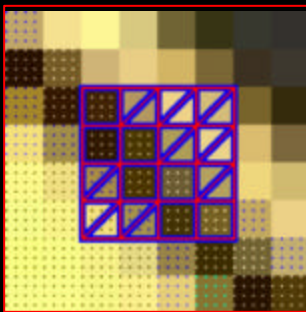
poking

affordance exploitation
(rolling)

object segmentation



edge catalog



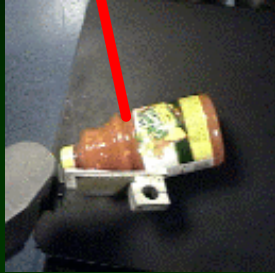
object detection
(recognition, localization,
contact-free segmentation)



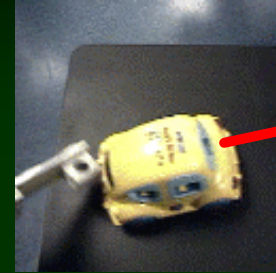
manipulator detection
(robot, human)



objects roll in different ways



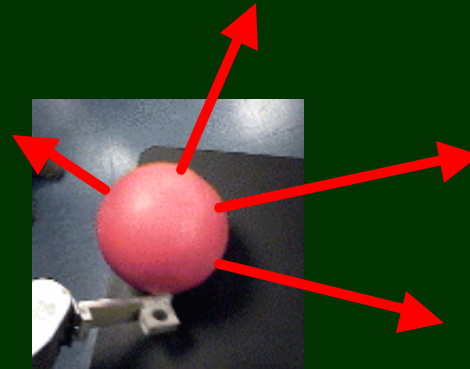
a bottle
it rolls along its side



a toy car
it rolls forward

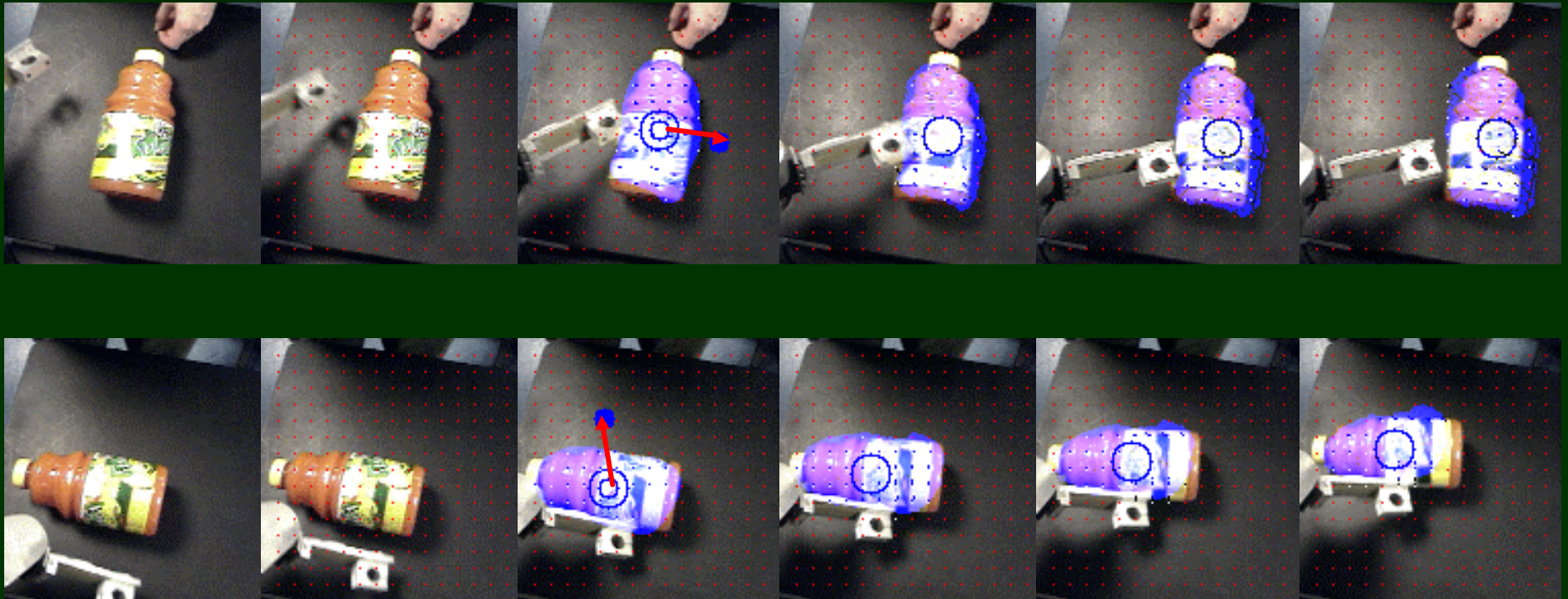


a toy cube
it doesn't roll easily



a ball
it rolls in
any direction

affordance exploitation

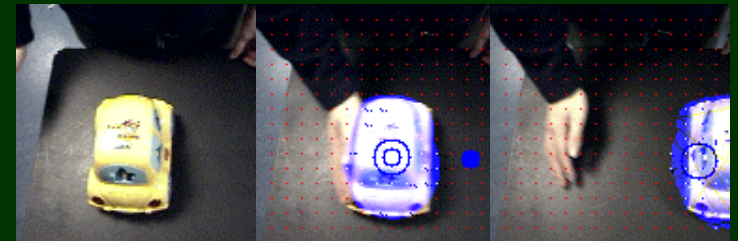
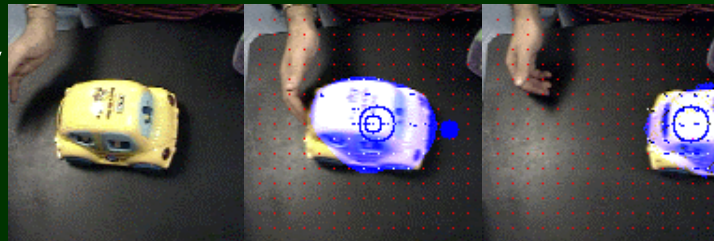


mimicry test

Invoking the object's natural rolling affordance

Going against the object's natural rolling affordance

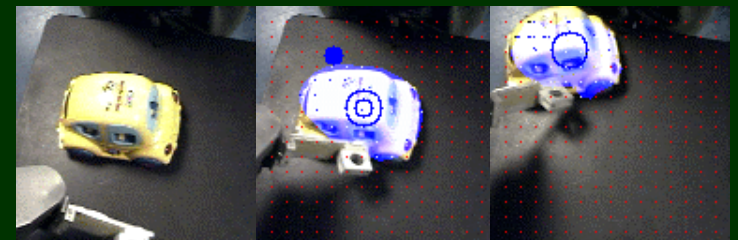
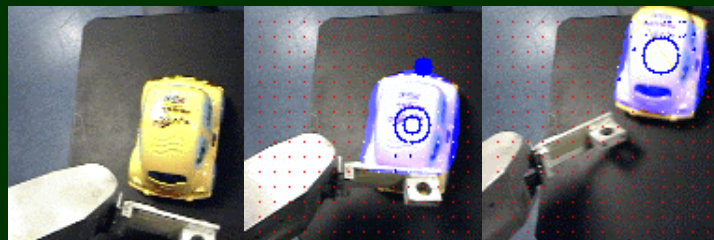
Demonstration by human



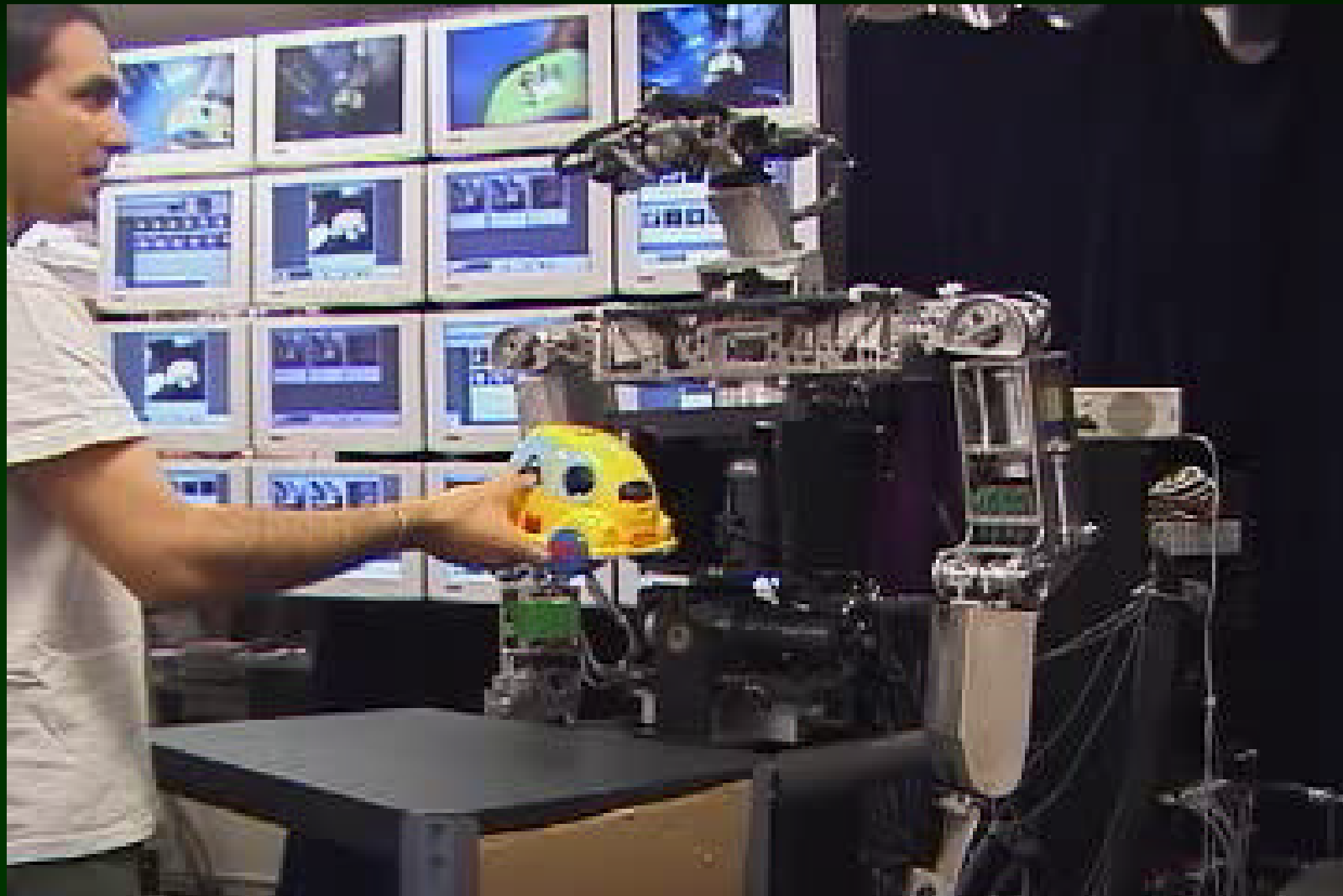
Mimicry in similar situation



Mimicry when object is rotated



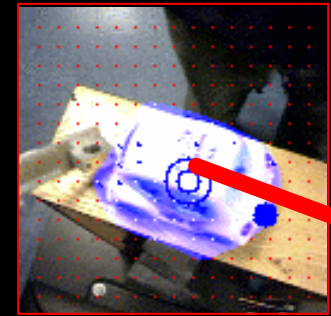
mimicry test



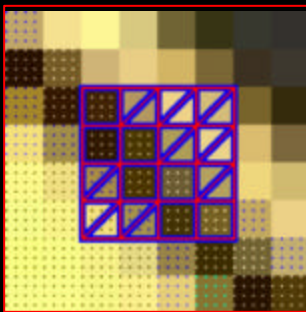
poking

affordance exploitation
(rolling)

object segmentation



edge catalog



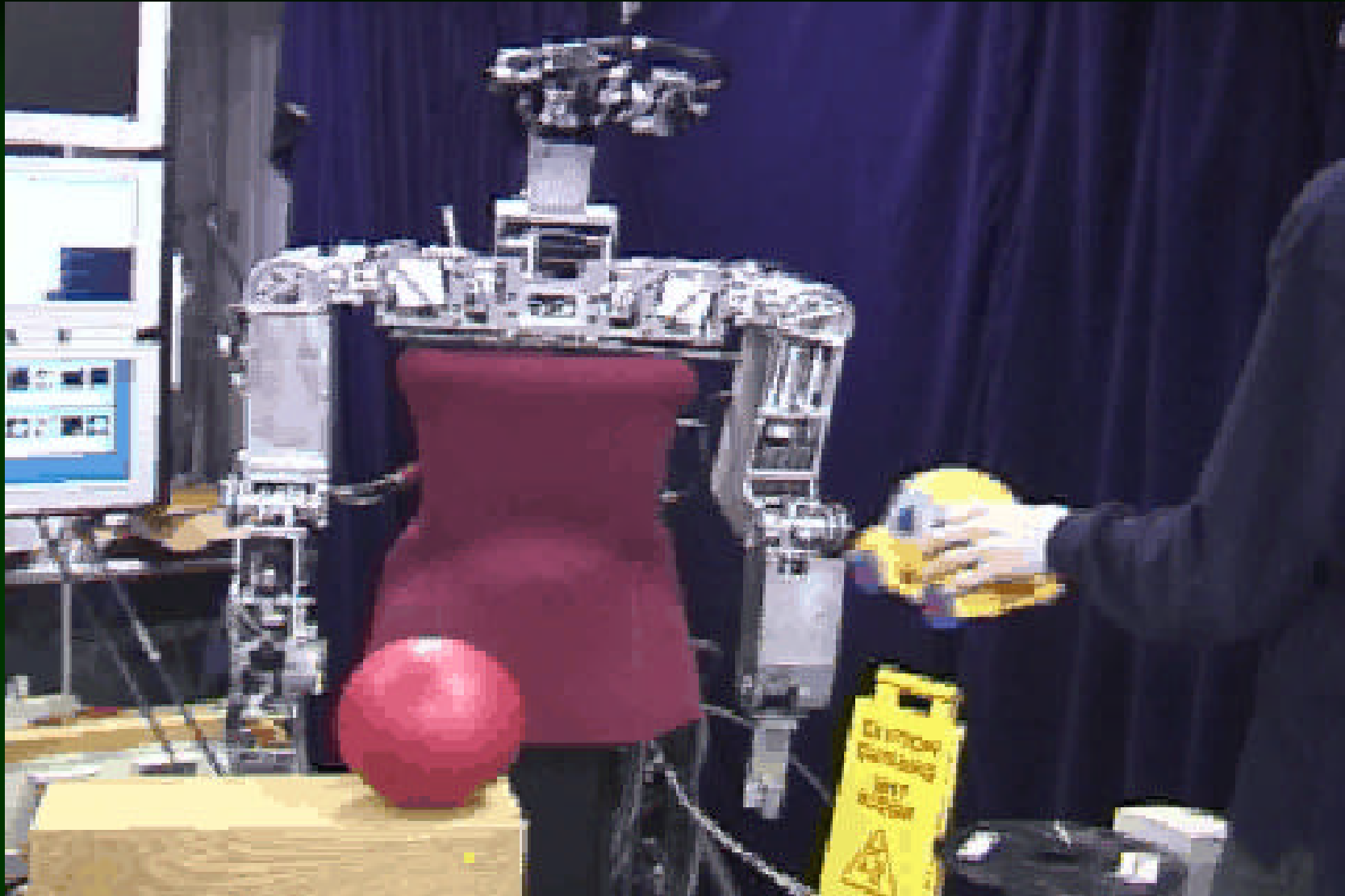
object detection
(recognition, localization,
contact-free segmentation)



manipulator detection
(robot, human)

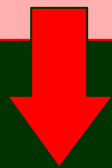


referring to objects

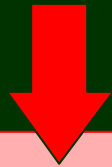


beyond objects

familiar activities



use constraint of familiar activity to discover unfamiliar entity used within it



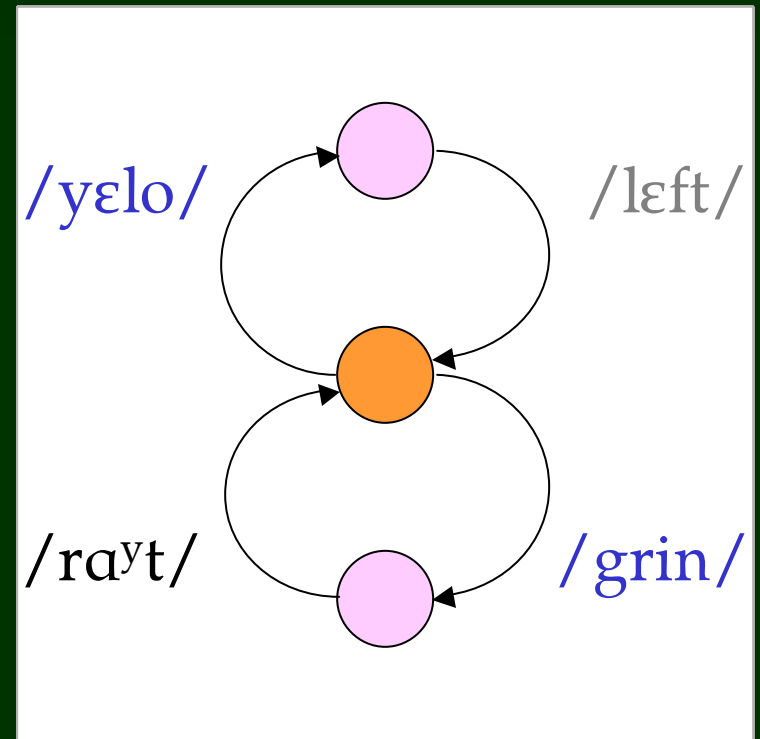
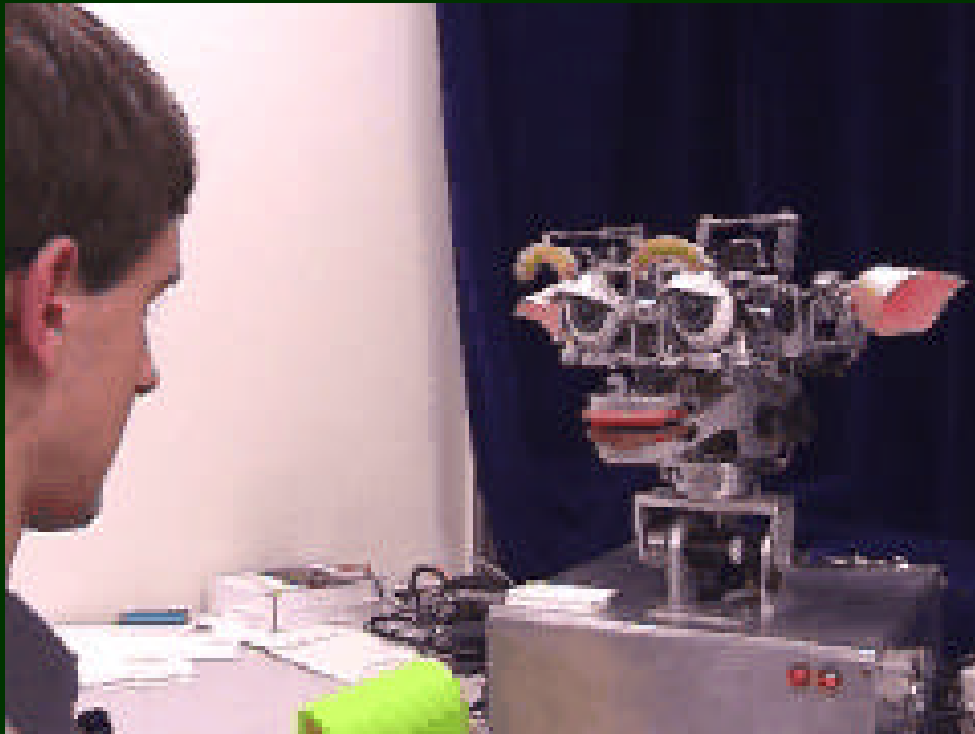
familiar entities (objects, actors, properties, ...)



reveal the structure of unfamiliar activities by tracking familiar entities into and through them



learning activity structure



conclusions

active segmentation	(through contact)
appearance catalog	(for oriented features)
open object recognition	for correction, enrollment
affordance recognition	(for rolling)
open speech recognition	(for isolated words)
virtuous circle of development	learning about and through activity