

# **A System for General In-Hand Object Re-Orientation**

CoRL 2021 best paper

# Authors

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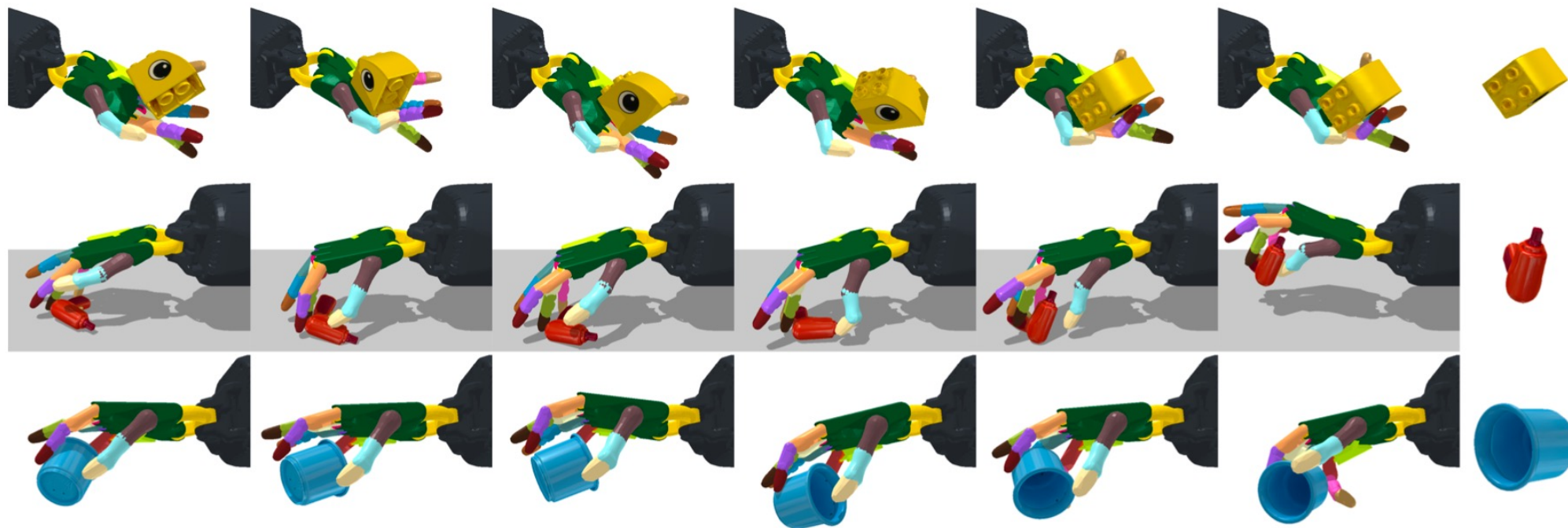
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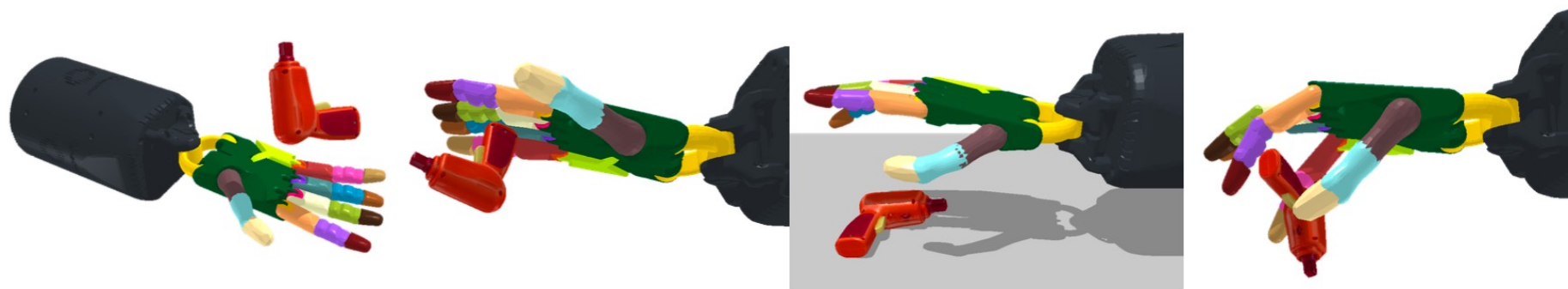
# Overview

- A model-free framework that learns to reorient objects of all kinds



# Contributions

- Hand facing upward and downward
- Zero-shot transfer on new objects
- Try vision-based observation

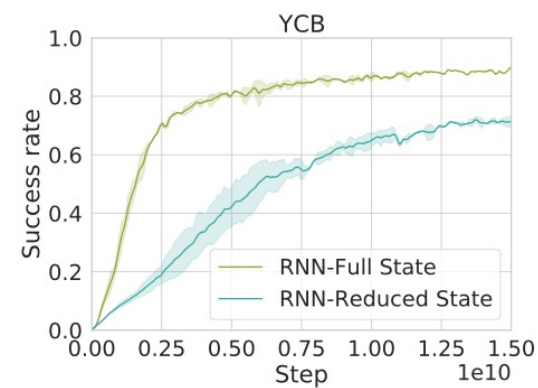
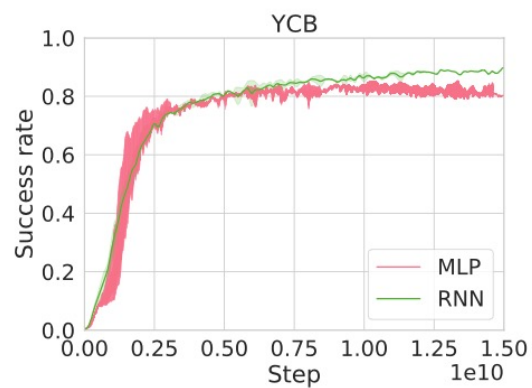
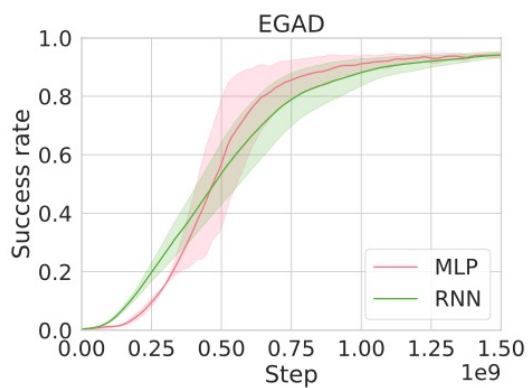


# Method

- Teacher-student Learning
- Gravity Curriculum
- Stable initialization

# Teacher Policy

- Full Observation
- MLP or RNN
- Dynamic randomization



# Teacher Policy

- Reward Function

$$r(s_t, a_t) = c_{\theta_1} \frac{1}{|\Delta\theta_t| + \epsilon_\theta} + c_{\theta_2} \mathbb{1}(|\Delta\theta_t| < \bar{\theta}) + c_3 \|a_t\|_2^2$$

# Student Policy

## Reduced Observation

Which can be obtained when in real world

## Vision or Non-vision

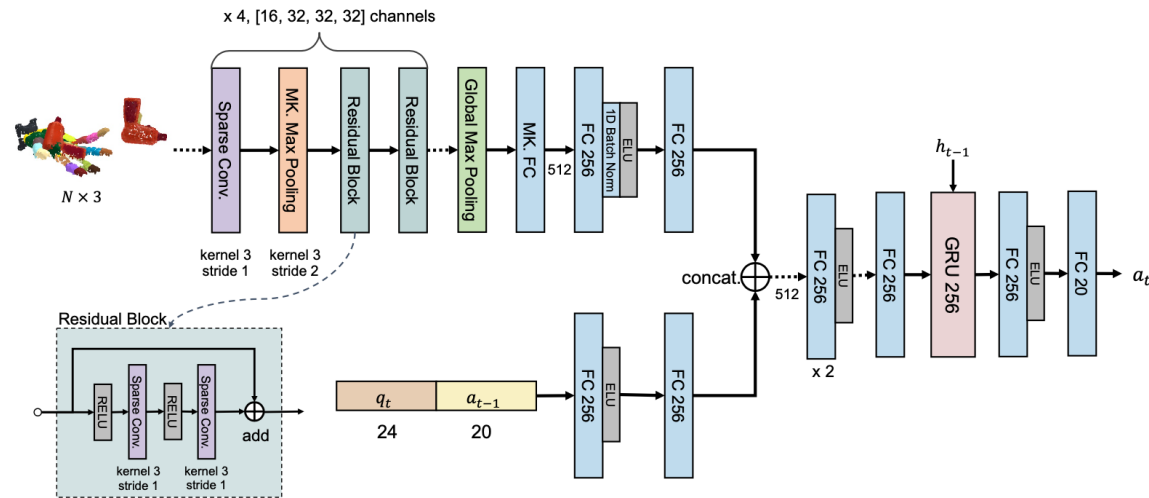
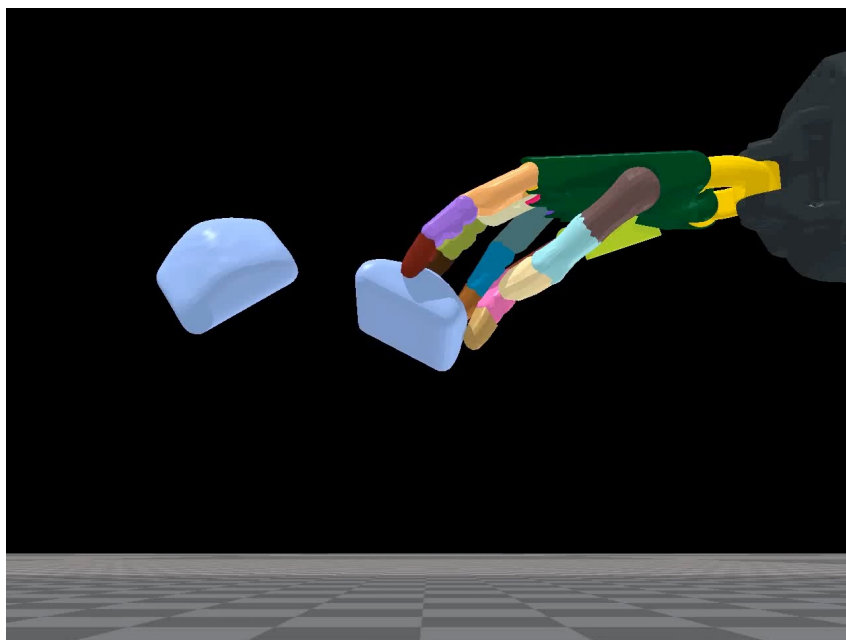


Figure 2: Visual policy architecture. MK stands for Minkowski Engine.  $q_t$  is the joint positions and  $a_t$  is the action at time step  $t$ .



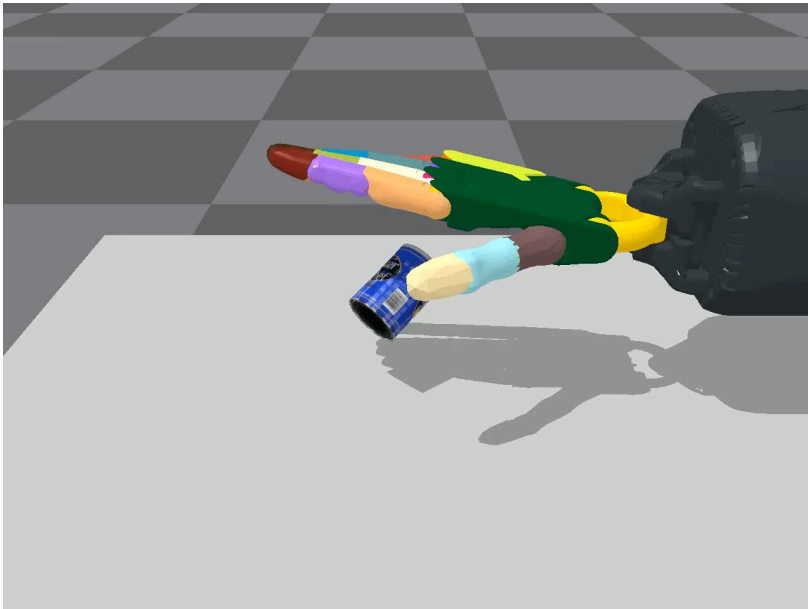
# Gravity Curriculum

- Hand facing downward and in air
- Gradually decrease  $g$



# Stable Initialization

- Reorient in air
- A separate RNN to lift objects



# Experiments

## ■ Dataset

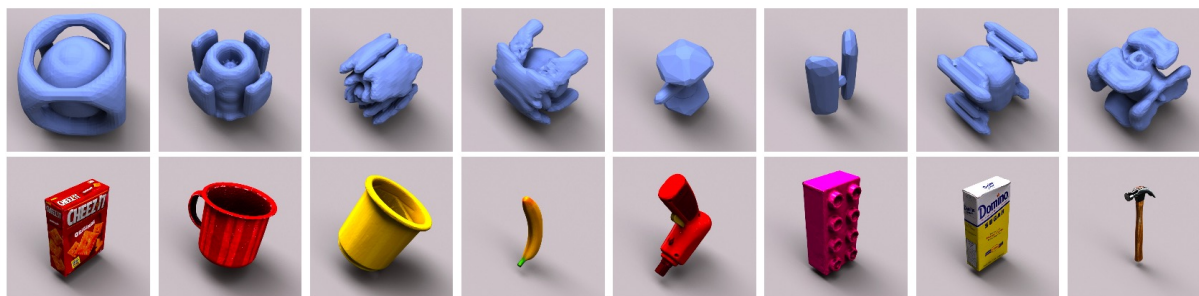


Figure B.2: First row: examples of EGAD objects. Second row: examples of YCB objects.

## ■ Facing upward and downward (with and without table)

## ■ Vision-based student policy

# Facing upward

## Results

Exp. ID	Dataset	State	Policy	1	2	3
				Train without DR		Train with DR
				Test without DR	Test with DR	Test with DR
B E	EGAD	Full state	RNN	$95.95 \pm 0.8$	$84.27 \pm 1.0$	$88.04 \pm 0.6$
		Reduced state	RNN→RNN	$91.96 \pm 1.5$	$78.30 \pm 1.2$	$80.29 \pm 0.9$
G J	YCB	Full state	RNN	$80.40 \pm 1.6$	$65.16 \pm 1.0$	$72.34 \pm 0.9$
		Reduced state	RNN→RNN	$81.04 \pm 0.5$	$64.93 \pm 0.2$	$65.86 \pm 0.7$

## Throw and Catch

## Failure

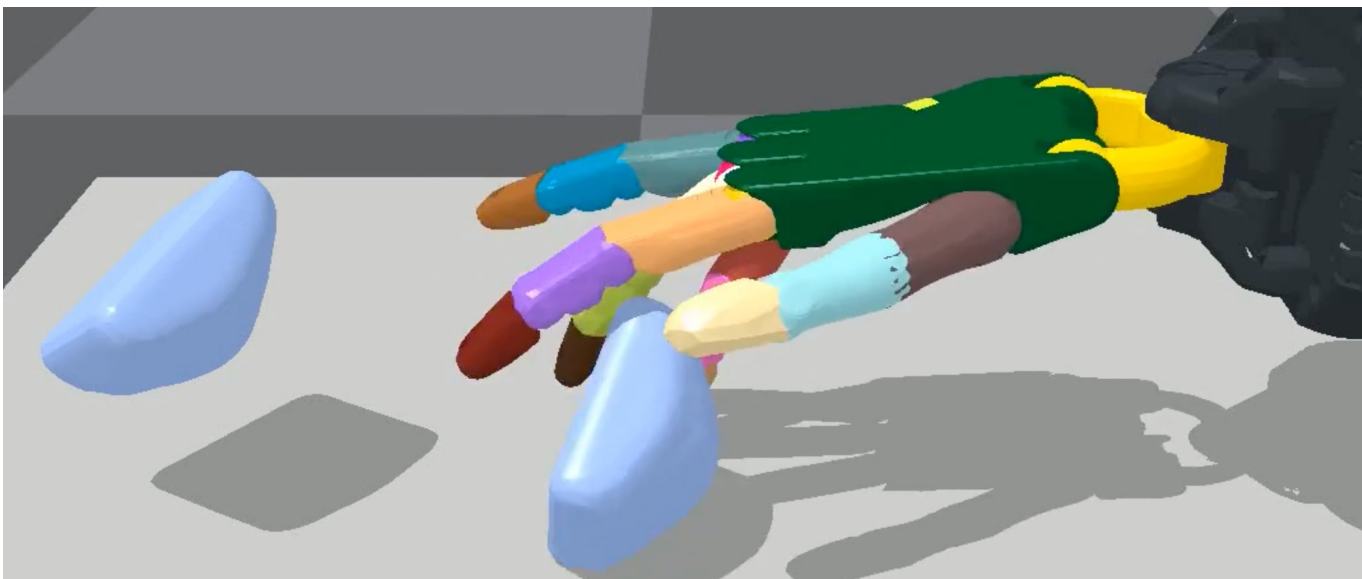


# Facing downward with table

## ■ Results

MLP policy for EGAD and YCB is  $95.31\% \pm 0.9\%$  and  $81.59\% \pm 0.7\%$

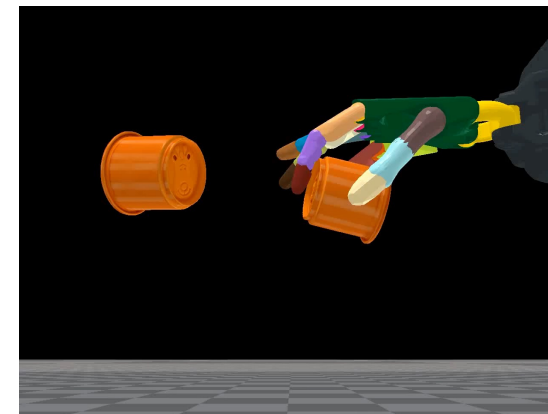
## ■ External Force



# Facing downward without table

## Results

Exp. ID	Dataset	State	Policy	1	2	3
				Train without DR		Finetune with DR
				Test without DR	Test with DR	Test with DR
K	EGAD	Full state	MLP	<b>84.29 ± 0.9</b>	<b>38.42 ± 1.5</b>	<b>71.44 ± 1.3</b>
L			RNN	82.27 ± 3.3	36.55 ± 1.4	67.44 ± 2.1
M		Reduced state	MLP→RNN	<b>77.05 ± 1.6</b>	29.22 ± 2.6	59.23 ± 2.3
N			RNN→RNN	74.10 ± 2.3	<b>37.01 ± 1.5</b>	<b>62.64 ± 2.9</b>
O	YCB	Full state	MLP	58.95 ± 2.0	26.04 ± 1.9	44.84 ± 1.3
P			RNN	52.81 ± 1.7	<b>26.22 ± 1.0</b>	40.44 ± 1.5
Q		RNN + $g$ -curr	<b>74.74 ± 1.2</b>	25.56 ± 2.9	<b>54.24 ± 1.4</b>	
R		MLP→RNN	46.76 ± 2.5	<b>25.49 ± 1.4</b>	34.14 ± 1.3	
S		Reduced state	RNN→RNN	45.22 ± 2.1	24.45 ± 1.2	31.63 ± 1.6
T		RNN + $g$ -curr→RNN	<b>67.33 ± 1.9</b>	19.77 ± 2.8	<b>48.58 ± 2.3</b>	



## Also Throw and Catch

# Zero-shot Transfer

## Results







	EGAD $\rightarrow$ YCB	YCB $\rightarrow$ EGAD
<b>U.FS</b>	$68.82 \pm 1.7$	$96.41 \pm 1.2$
<b>U.RS</b>	$59.64 \pm 1.8$	$96.38 \pm 1.3$
<b>D.FS</b>	$62.73 \pm 2.2$	$85.45 \pm 2.9$
<b>D.RS</b>	$55.30 \pm 1.3$	$77.91 \pm 2.1$

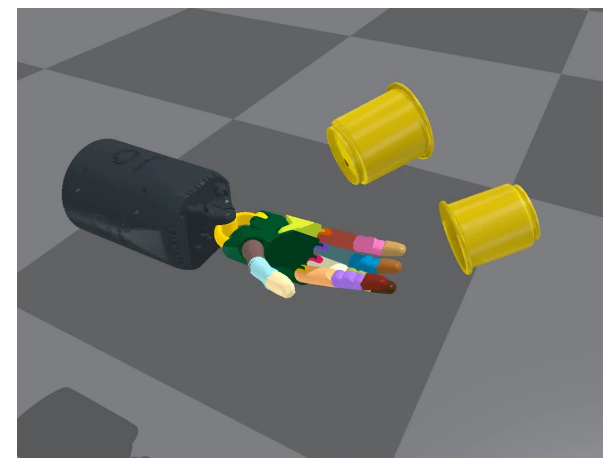
## Shape-agnostic

# Vision-based

## ■ Constraints

## ■ Results

	Object	Success rate (%)
	025_mug	$89.67 \pm 1.2$
	065-d_cups	$68.32 \pm 1.9$
	072-b_toy_airplane	$84.52 \pm 1.4$
	073-a_lego_duplo	$58.16 \pm 3.1$
	073-c_lego_duplo	$50.21 \pm 3.7$
	073-e_lego_duplo	$66.57 \pm 3.1$





# Comment

- Highly Dynamic
- Reduced State can be obtained?
- Shape-agnostic?