

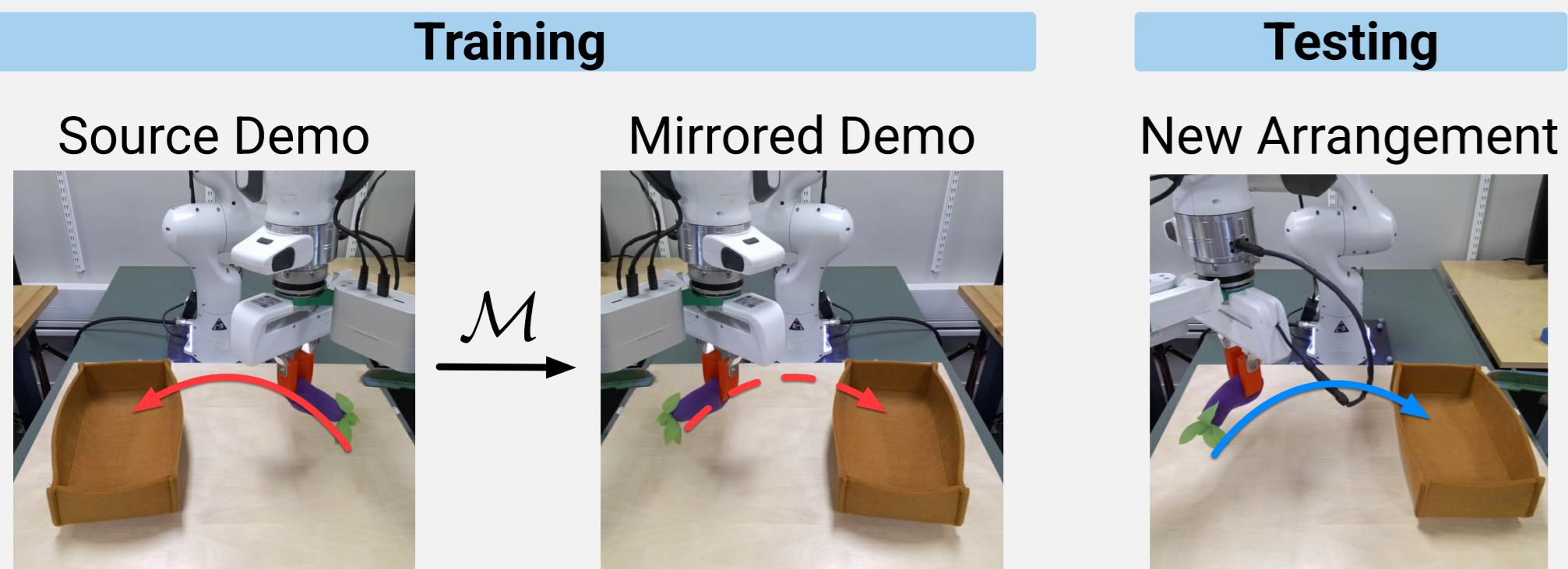
# MirrorDuo

## Reflection-Consistent Visuomotor Learning from Mirrored Demonstration Pairs

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### Get One Demo, Get One for Free



**MirrorDuo** captures a simple idea: moving left implies knowing how to move right. We turn this intuition into a framework for image-based imitation learning in the SE(3) state-action space.

#### - Workspace Generalization

**One-sided** demos can be mirrored to cover the opposite side, effectively expanding the workspace.

#### - Data Efficiency

For **both-sides** demonstrations, MirrorDuo boosts data efficiency, fewer demos are needed to span the same workspace.

### How Does MirrorDuo Work?

**Base formulation** mirrors end-effector poses (and actions)  $\mathbf{X}_H$  in the camera frame  $\mathbf{X}_C$  using a reflection operator  $\mathbf{E} = \text{diag}([-1, 1, 1, 1])$ . Mapping back to the world frame gives  $\mathbf{X}_H^* = \mathbf{X}_C \mathbf{E} \mathbf{X}_C^{-1} \mathbf{X}_H \mathbf{E}$ .

#### Remove Camera Extrinsics

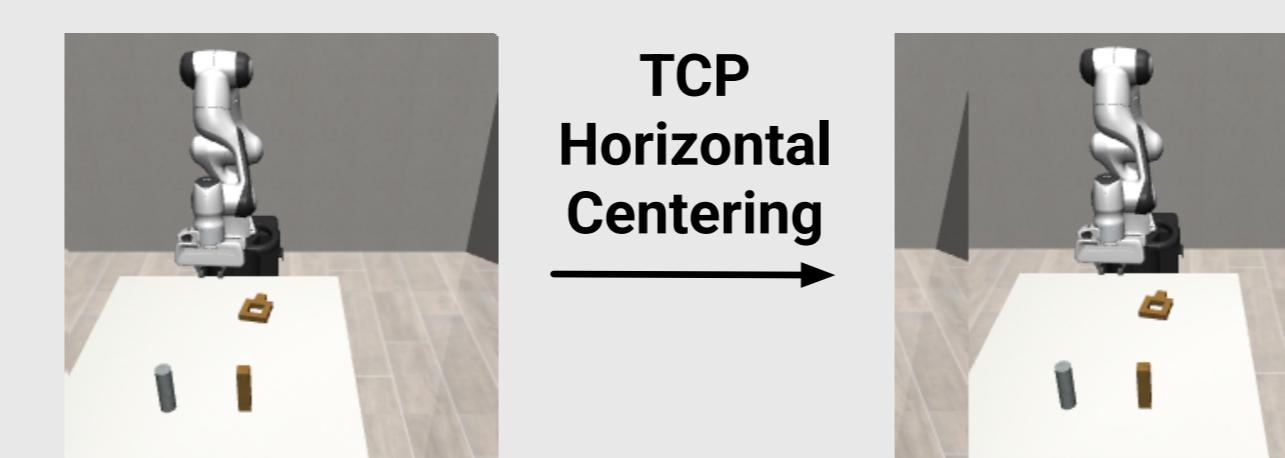
Absolute Pose  $\rightarrow$  Delta Poses

$$\Delta \mathbf{X}_{H_t} = \mathbf{X}_{H_0}^{-1} \mathbf{X}_{H_t}$$

Relative Poses unchanged

$$\delta \mathbf{X}_{H_t} = \mathbf{X}_{H_0}^{-1} \mathbf{X}_{H_t}$$

#### Center TCP to the Flipping Axis



#### Align Initial Rotation to the World Frame

Ensure the initial rotation remains unchanged under mirroring:

$$\mathbf{R}^* = \mathbf{E} \mathbf{R} \mathbf{E}.$$

These steps enable the policy to learn trajectories that are immediately deployable in the mirrored setup, **starting from the same initial pose**.

### How Effective Is MirrorDuo?

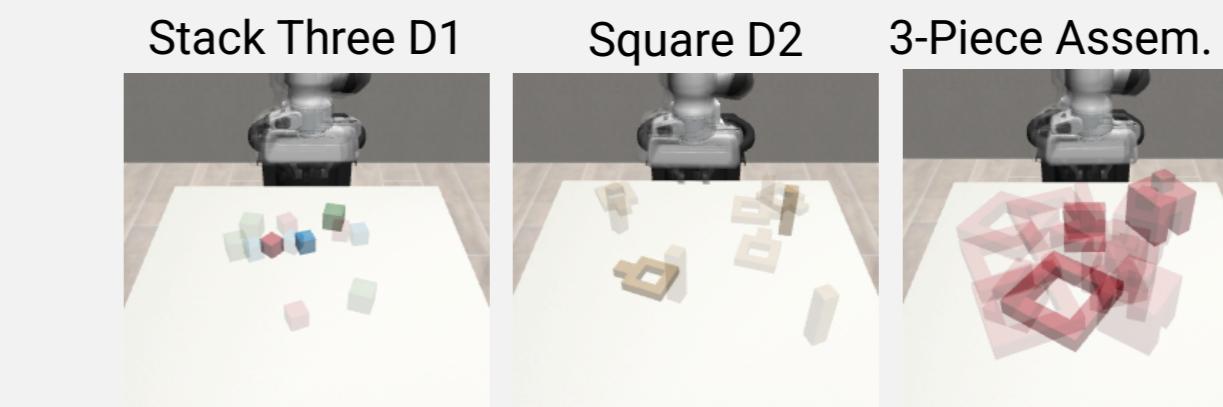


Table.1 Average **absolute success rate improvement** over three tasks, relative to each baseline, for both absolute and relative actions. ( $\mathcal{M}, \mathcal{O}, \mathcal{P}$ ) denote MirrorAug, Random Overlay, and Pretrained Encoder.

Method	Low Demos	Medium Demos	High Demos
MirrorDiff. + $\mathcal{O}$	13.1	13.3	3.2
DiffPo + $\mathcal{M}, \mathcal{O}, \mathcal{P}$	17.8	15.3	2.2
BC-RNN + $\mathcal{M}, \mathcal{O}, \mathcal{P}$	0.3	2.7	15.6
<b>Relative</b>			
MirrorDiff. + $\mathcal{O}$	13.6	20.7	17.5
DiffPo + $\mathcal{M}, \mathcal{O}, \mathcal{P}$	10.6	14.4	9.1
BC-RNN + $\mathcal{M}, \mathcal{O}, \mathcal{P}$	6.1	9.3	12.0

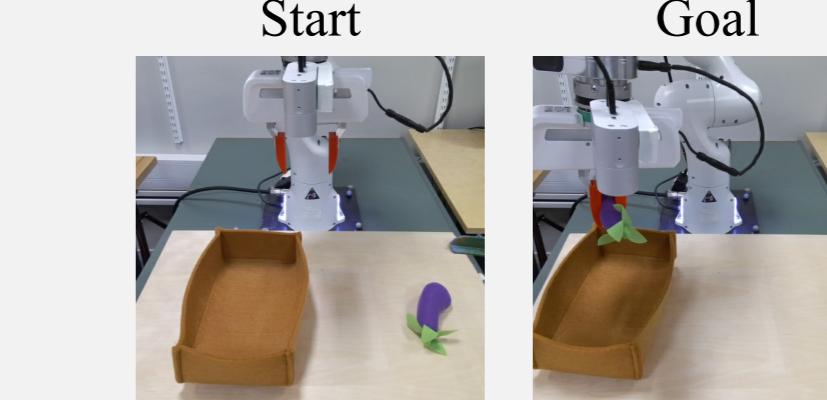


Table.2 Generalization to Mirrored Setup. Success rates (out of 30 trials) in original and mirrored setups. M-demos denote demonstrations in the mirrored setup.

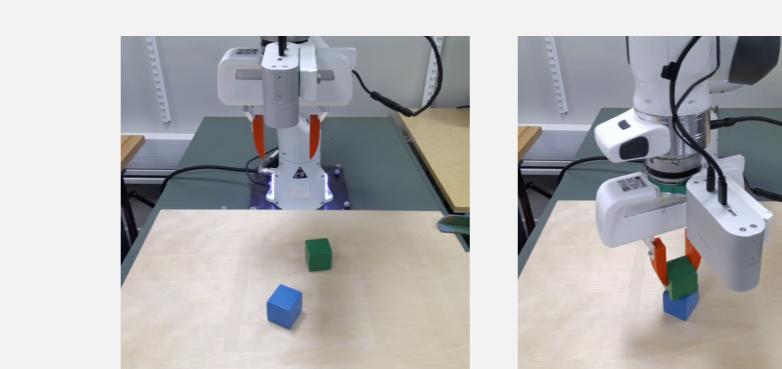
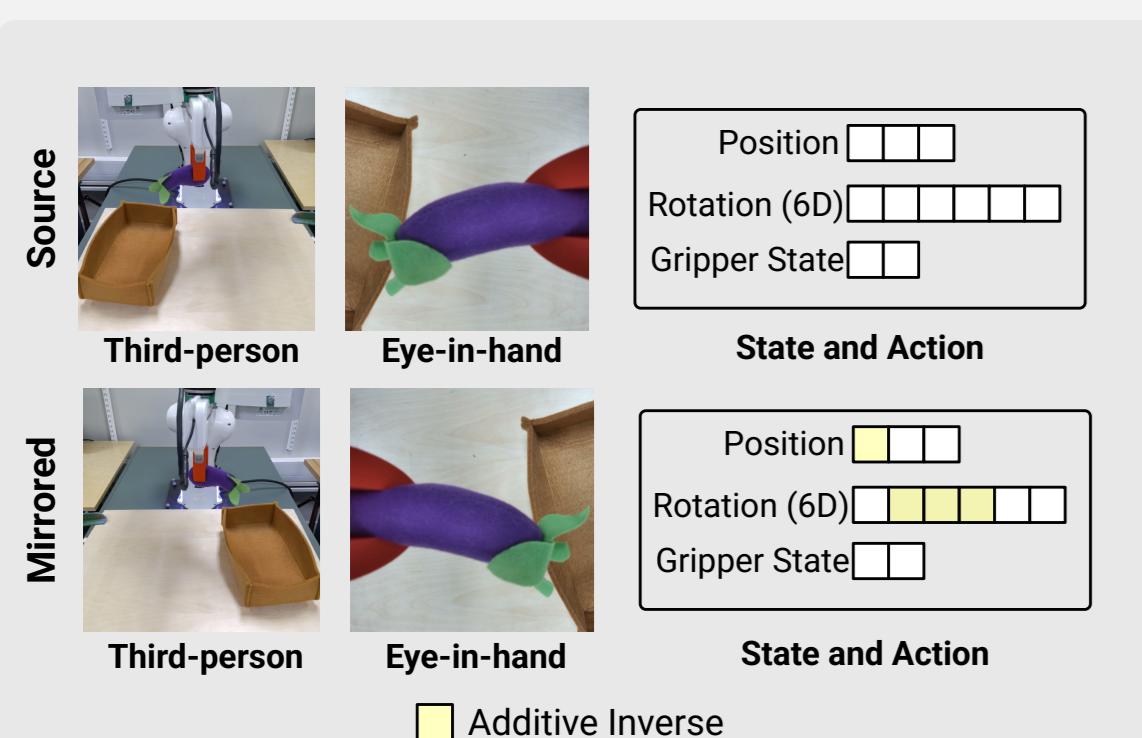


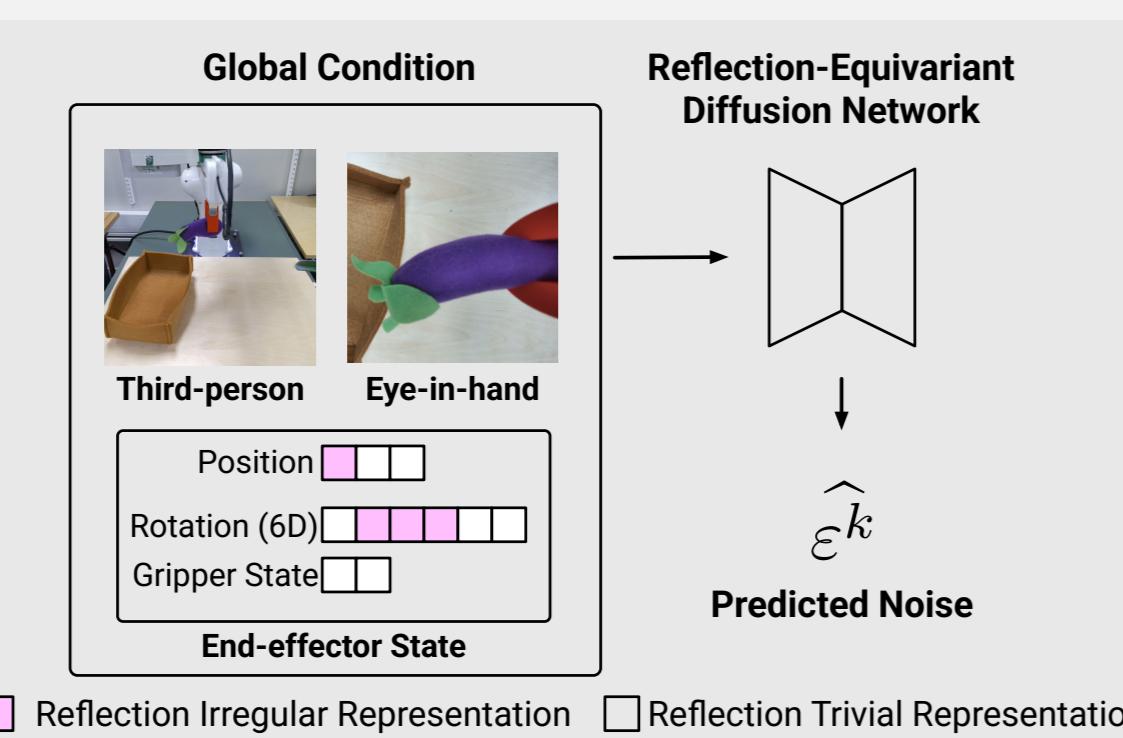
Table.3 In-domain Data Efficiency. Success rates (30-trial average) on block stacking

### Dual Realization



#### Data Augmentation (MirrorAug)

Synthesize mirrored samples (Image, Proprioception, Action)



#### Reflection-Equivariant Diffusion Policy (MirrorDiffusion)

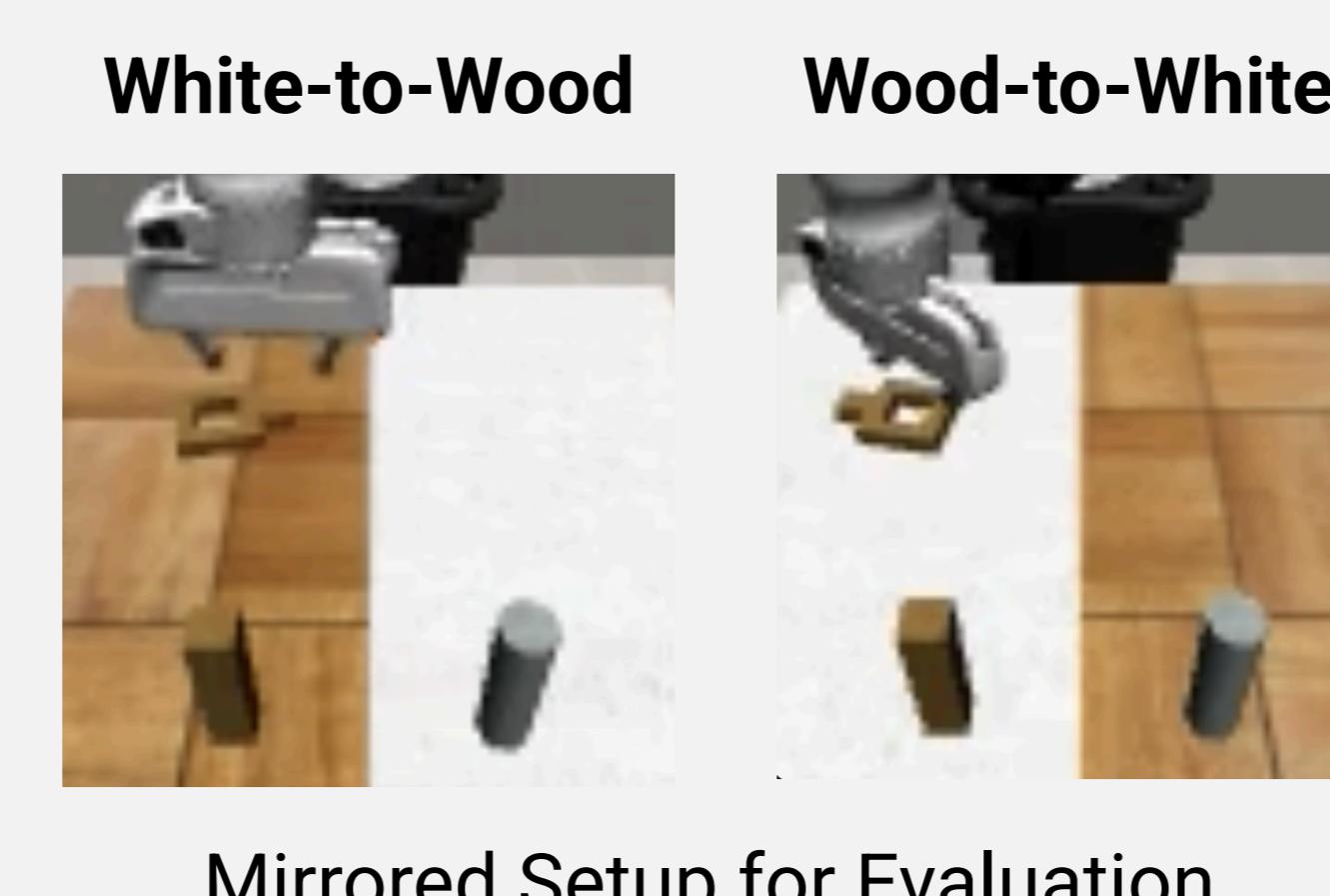
Use reflection-symmetry as inductive bias for Policy

### Visual Asymmetry as Distractors

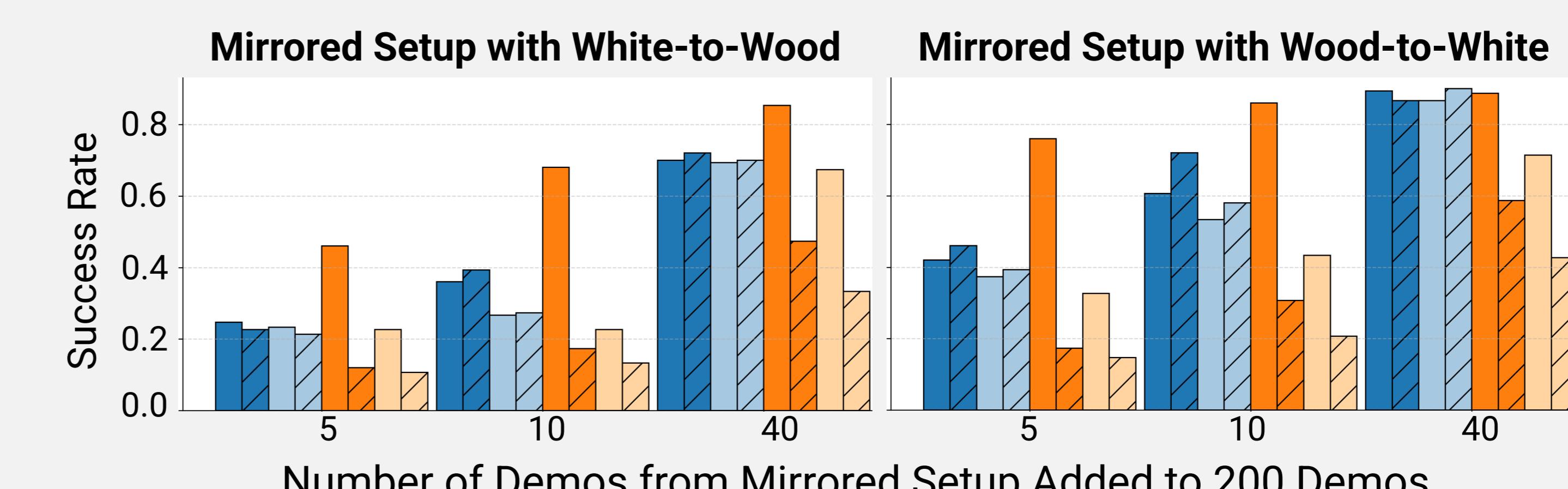
#### Visual asymmetry handling is closely tied to robustness against backgrounds, distractors, and lighting variations.

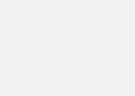
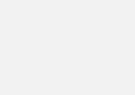
Even simple robustness techniques, such as finetuning ImageNet-pretrained encoders or applying Random Overlay, yield significant gains.

With enhanced visual generalization, MirrorDuo greatly reduces the real demos needed for transfer to the opposite side.



Mirrored Setup for Evaluation



**M** MirrorDuo as Augmentation  
**O** Random Overlay  
**P** Pretrained ResNet18  
 **M**  
 **Diffusion Policy + M, O, P**  
 **Diffusion Policy + M**  
 **Mirror-Equi Diffusion + M, O**  
 **Mirror-Equi Diffusion + M**