

Toward Understanding Compositional Generalization in Object-Oriented World Modeling

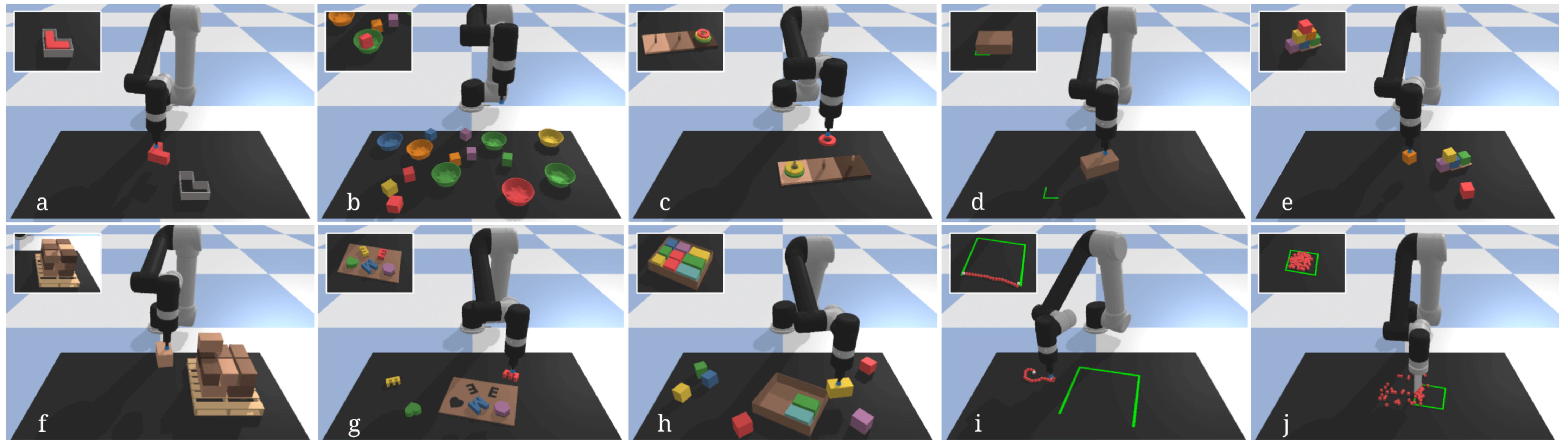
Linfeng Zhao, Lingzhi Kong, Robin Walters, Lawson Wong
Khoury College of Computer Sciences, Northeastern University

International Conference on Machine Learning, 2022

Motivation

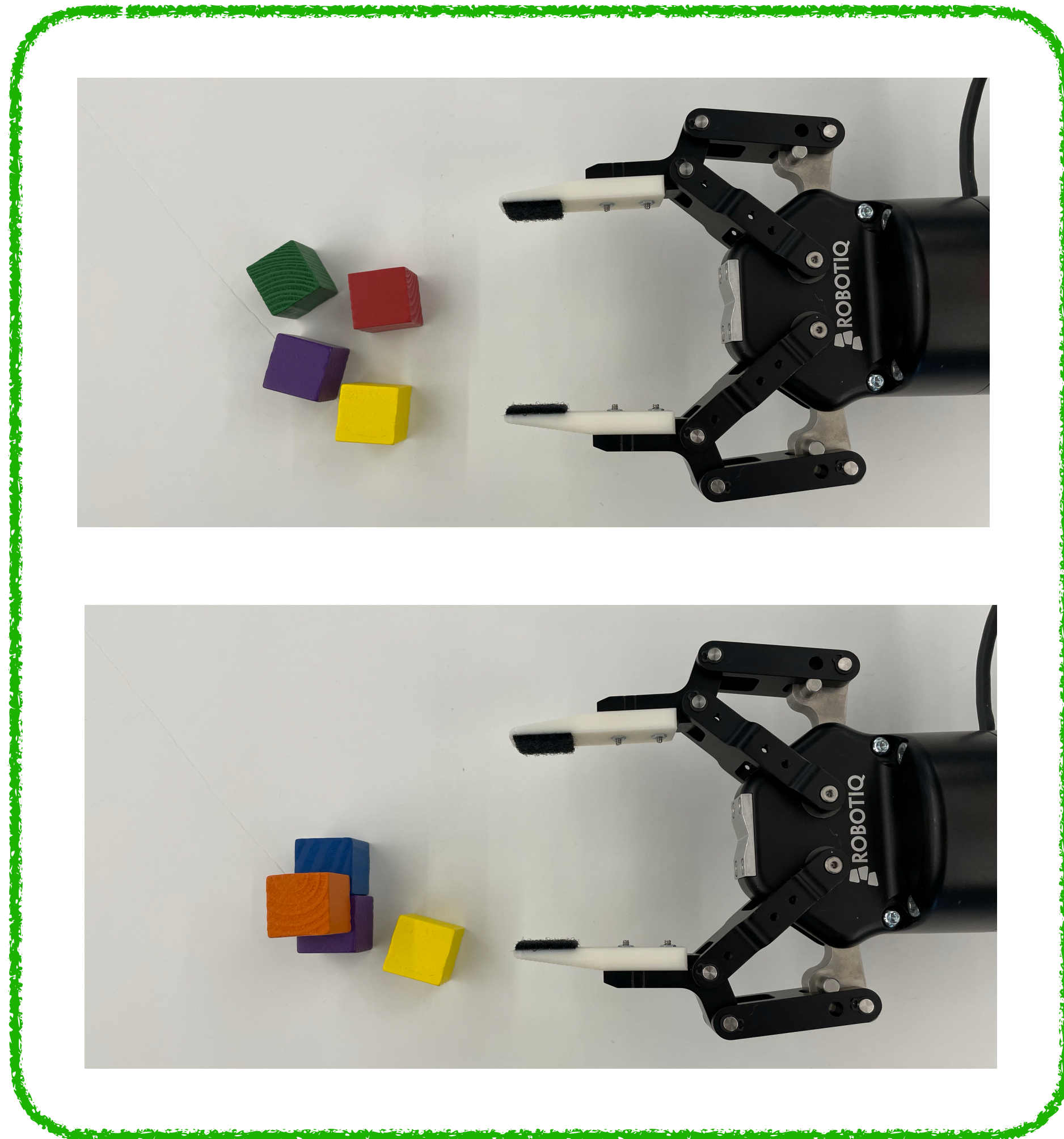
Motivation: Robotic Manipulation

Example: Robotic Manipulation

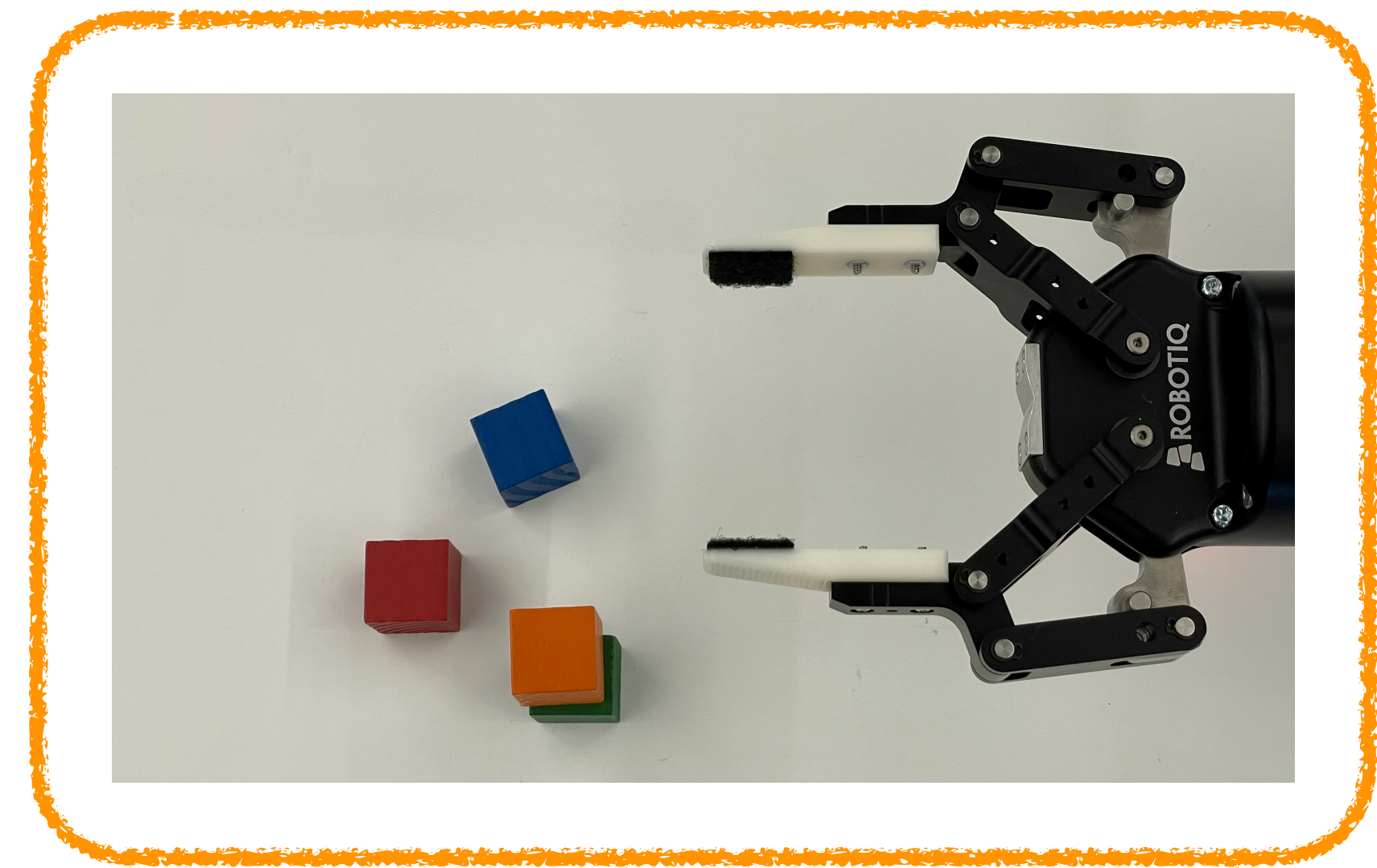
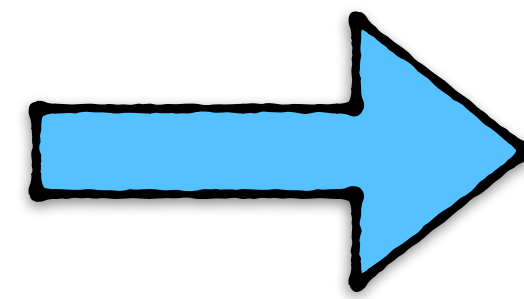


Credit: Ravens

Motivation: Robotic Manipulation



Train



Generalization



Research Questions

Compositional Generalization (CG)

- How to **measure** **compositional generalization** (in world modeling)?
- Can we **guarantee** when **compositional generalization** is achieved?
- If possible, how can its **implementation** be **efficient**?

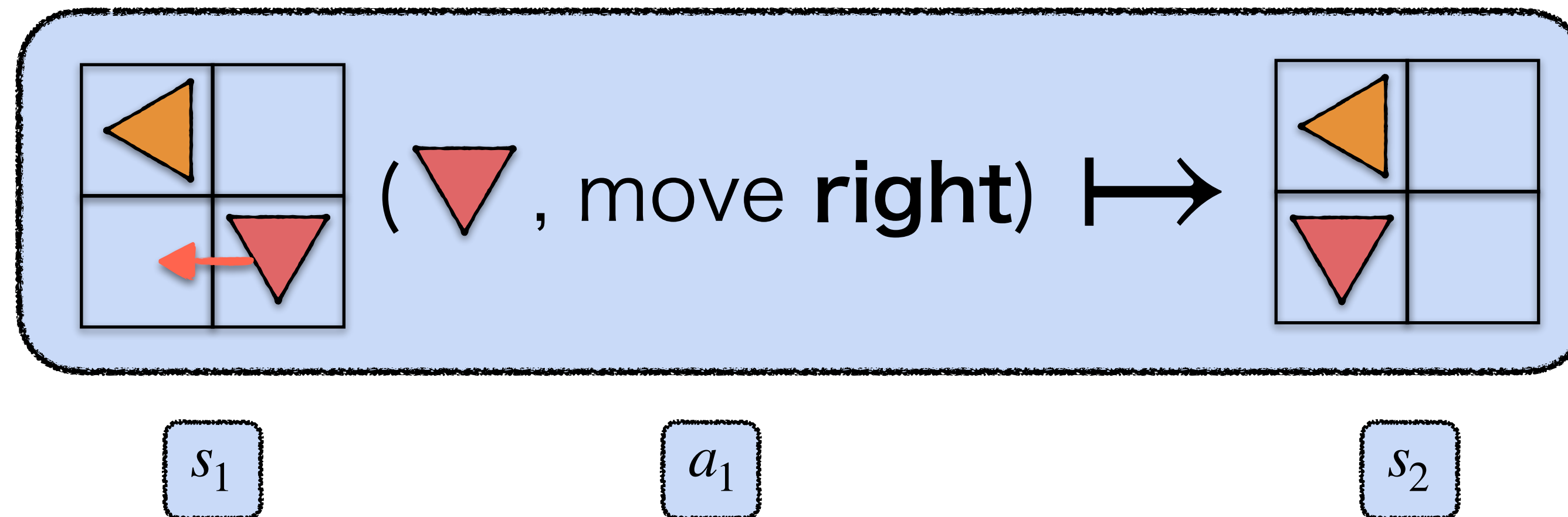
Outline

- Motivation
- Setup
- Defining Compositional Generalization
- Implementing Compositional Generalization
- Solving Binding Issue for Compositional Generalization
- Results

Setup

World Modeling

Learn a transition model $T : \mathcal{S} \times \mathcal{A} \rightarrow \mathcal{S}$



Actions are factorized by objects and relative to orientation

Object Library

Object Library \mathbb{L}
 "Vocabulary"

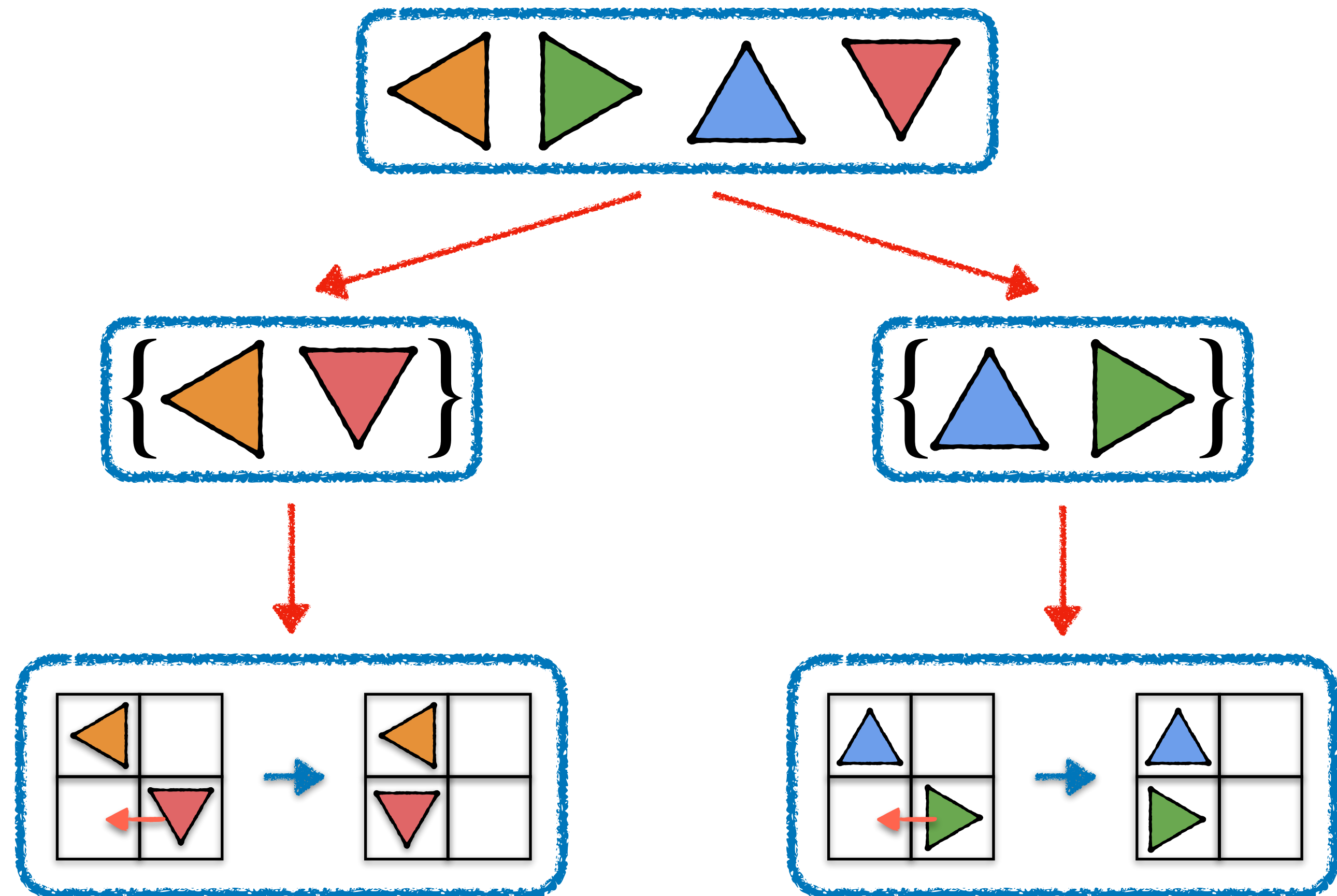
$$N \triangleq |\mathbb{L}| = 4$$

Scene Sets $\mathbb{O}_i \subset \mathbb{L}$
 (Ordered) "Sentences"

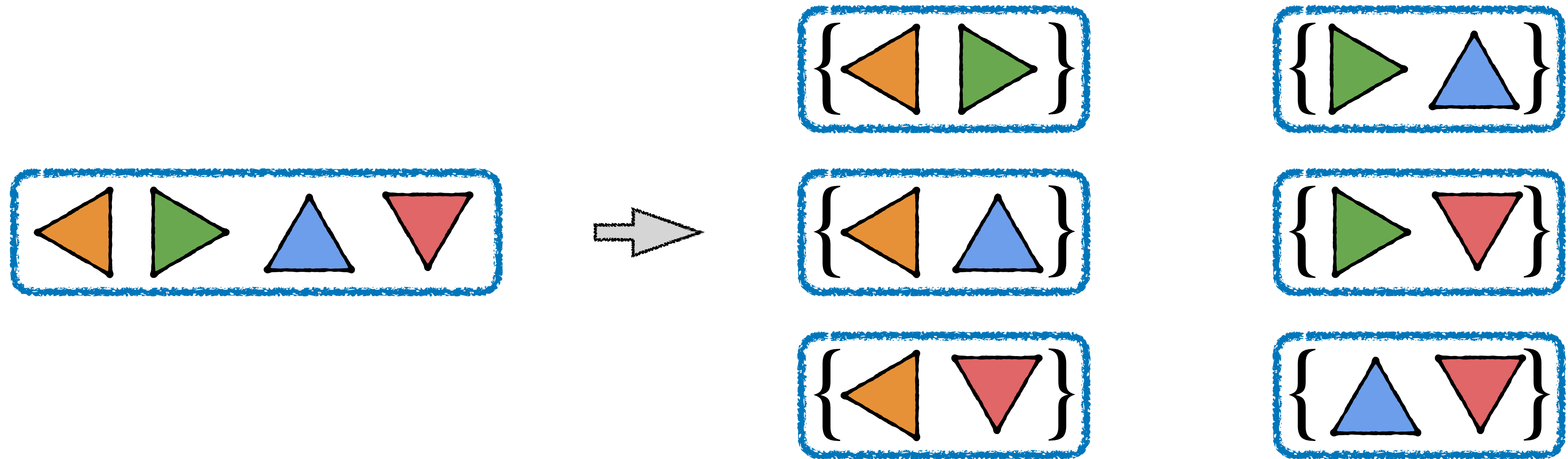
$$K \triangleq |\mathbb{O}| = 2$$

Scene MDPs $\mathcal{M}_{\mathbb{O}_i}$

Generated by \mathbb{O}



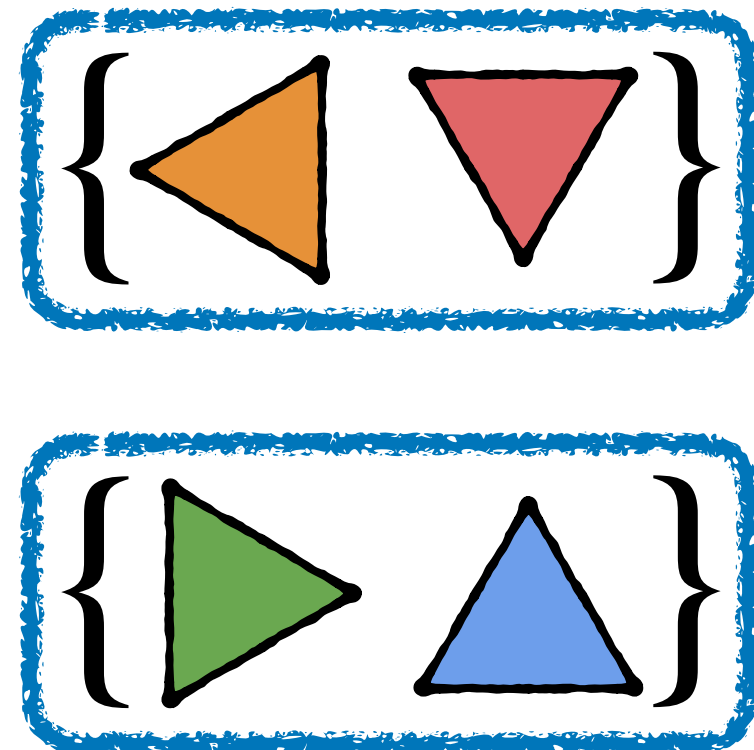
Object Library



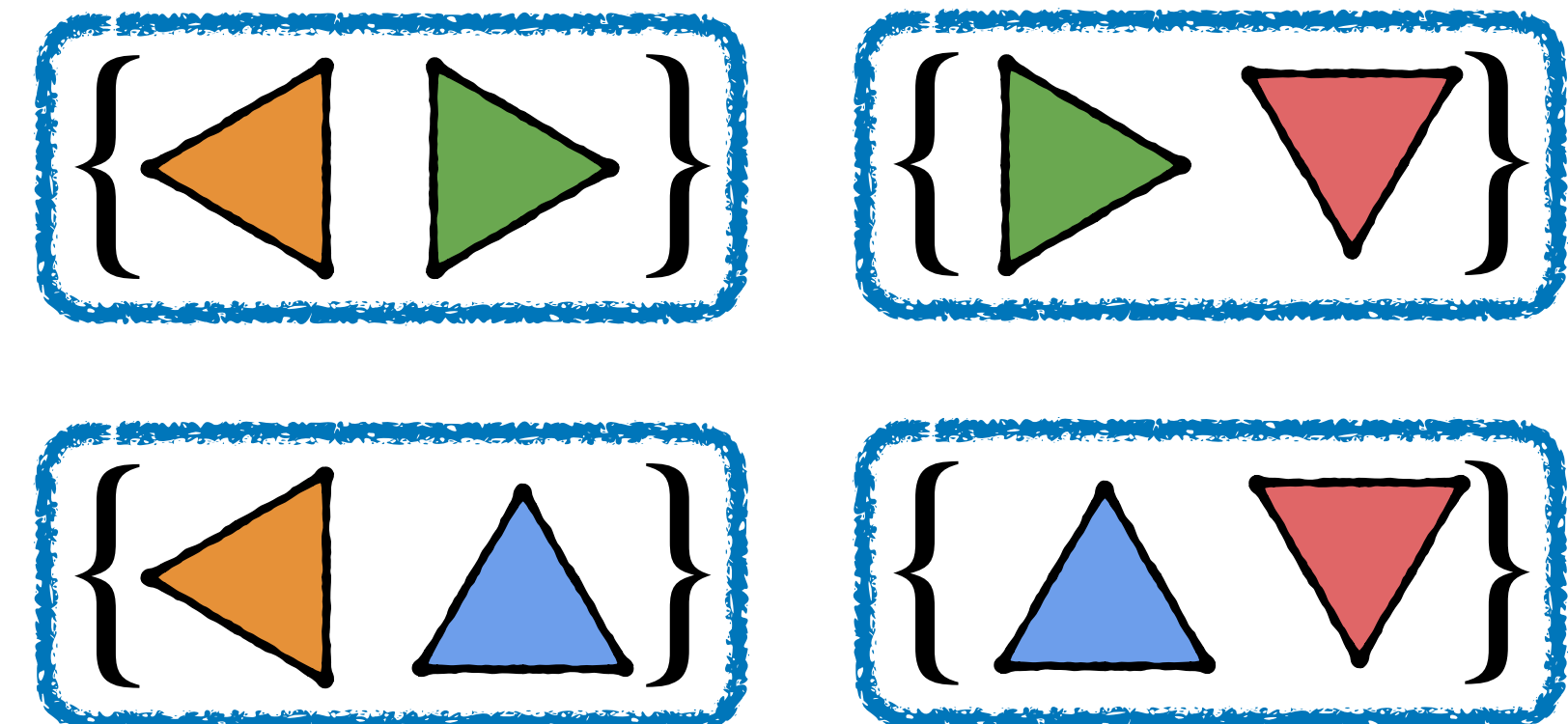
All $\binom{N}{K}$ combinations

Object Library

Training



Generalization



Learn a transition model $T : \mathcal{S} \times \mathcal{A} \rightarrow \mathcal{S}$

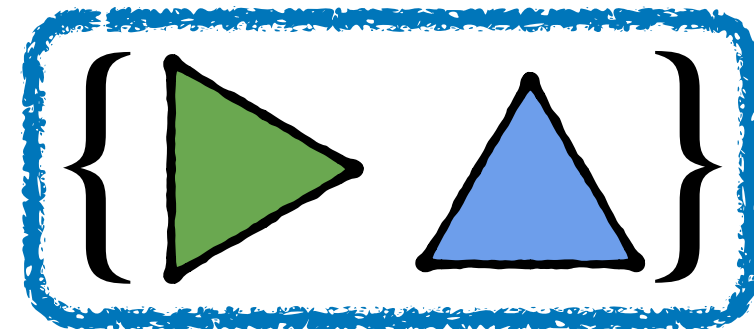
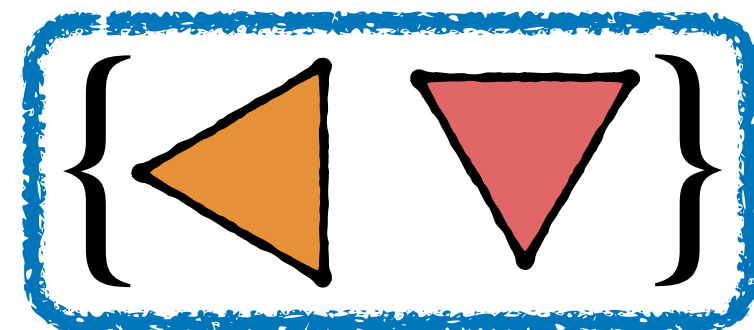
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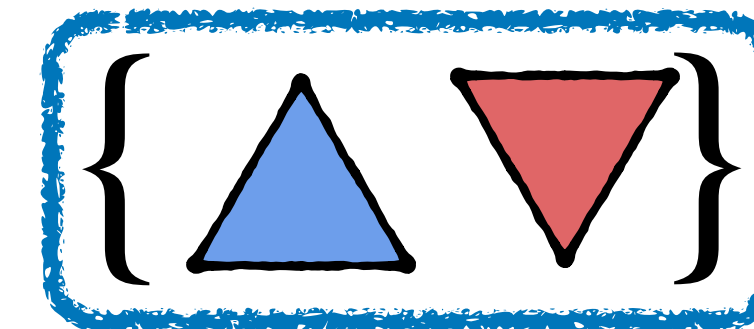
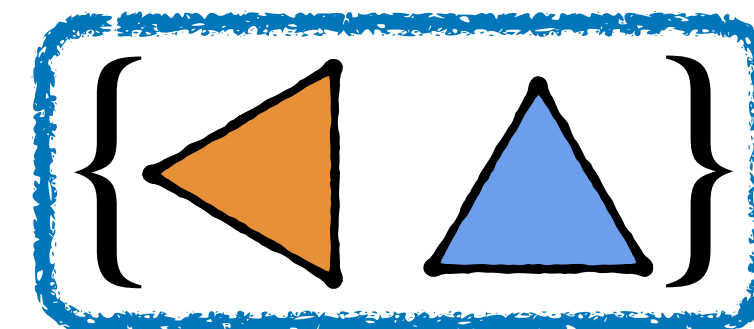
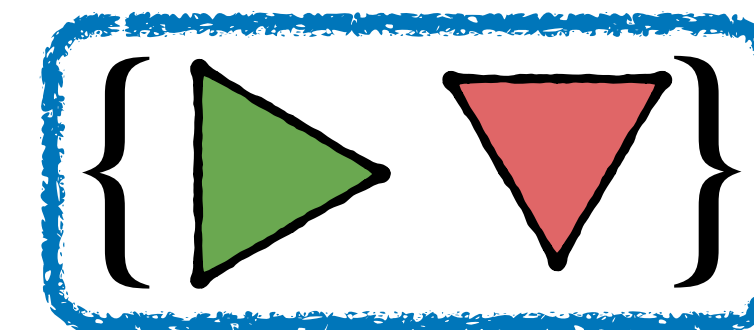
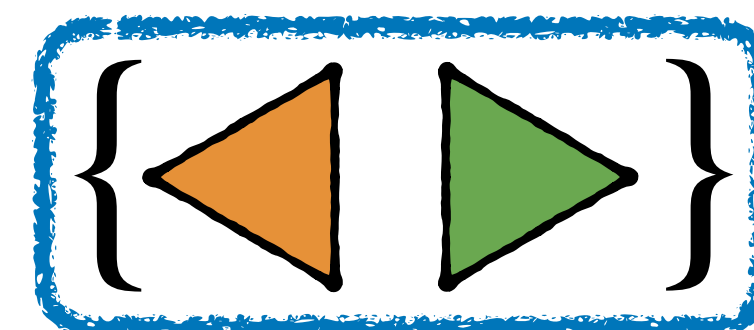
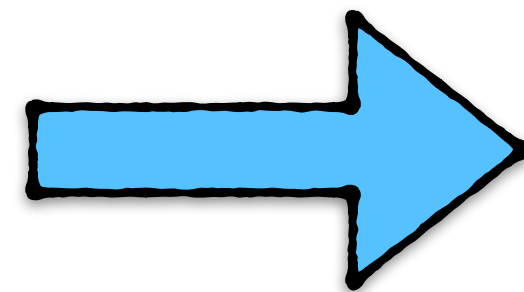
Defining Compositional Generalization

Motivation

How to quantitatively define and measure compositional generalization?



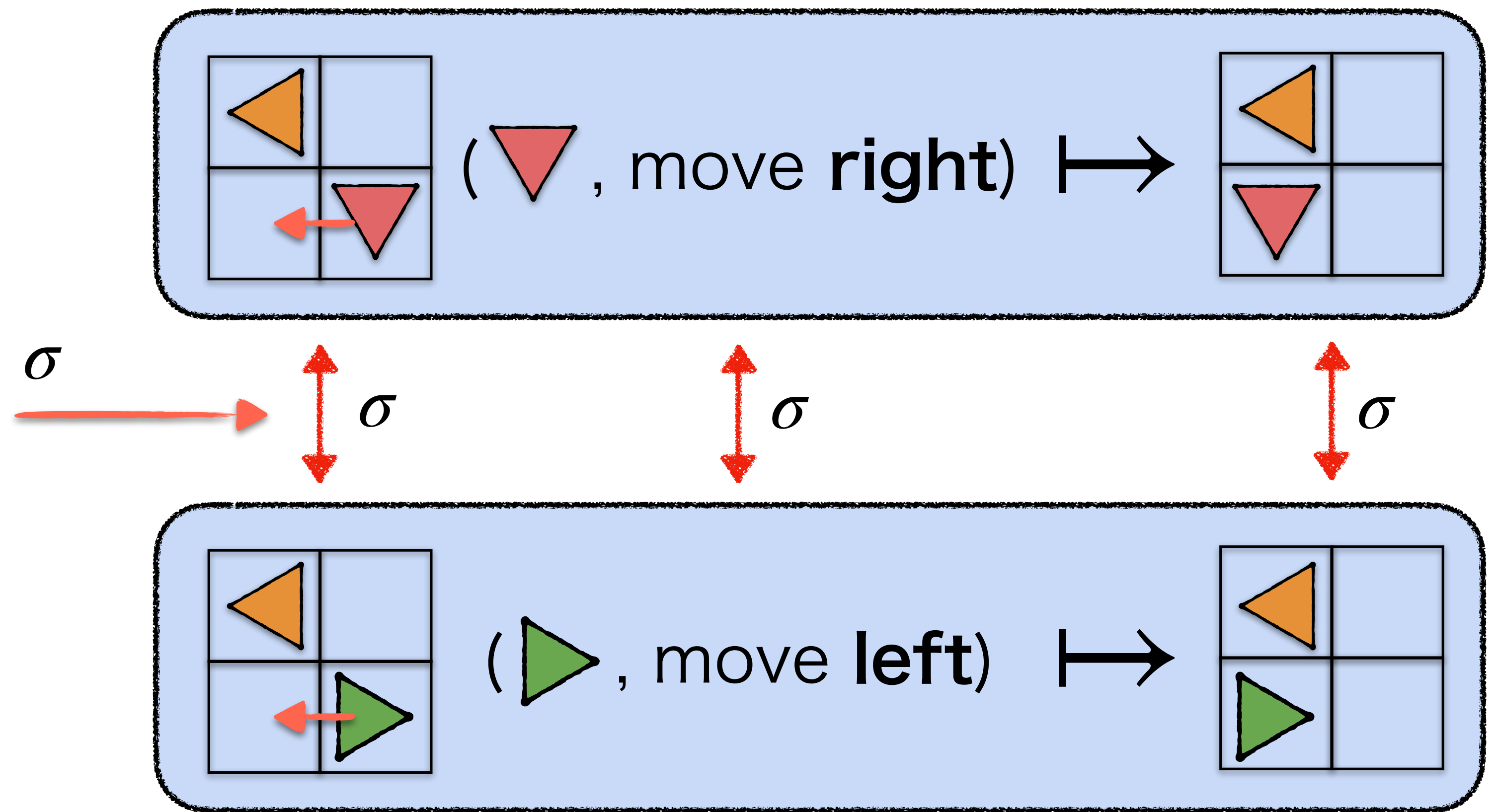
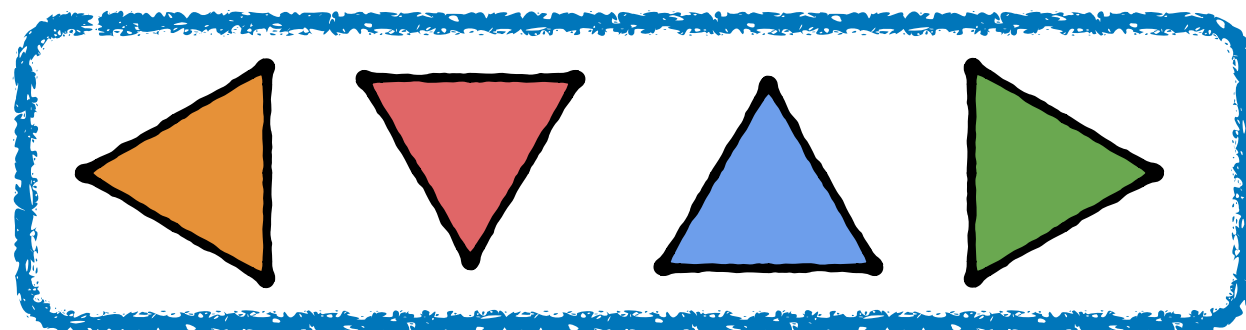
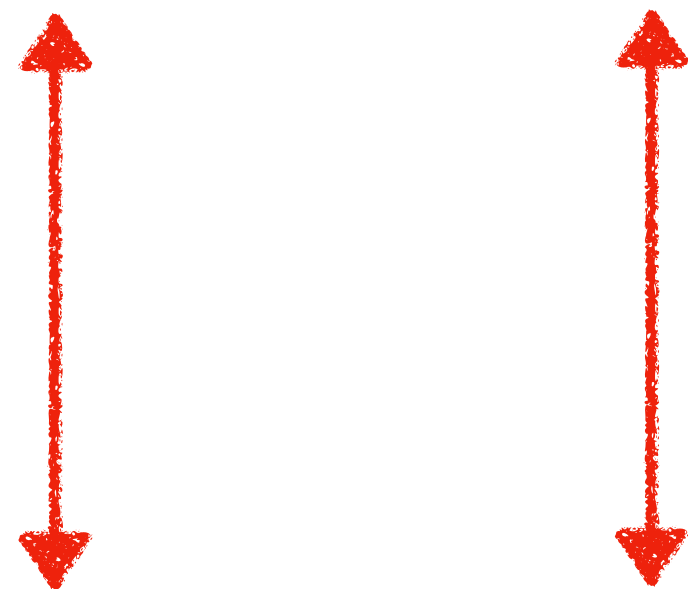
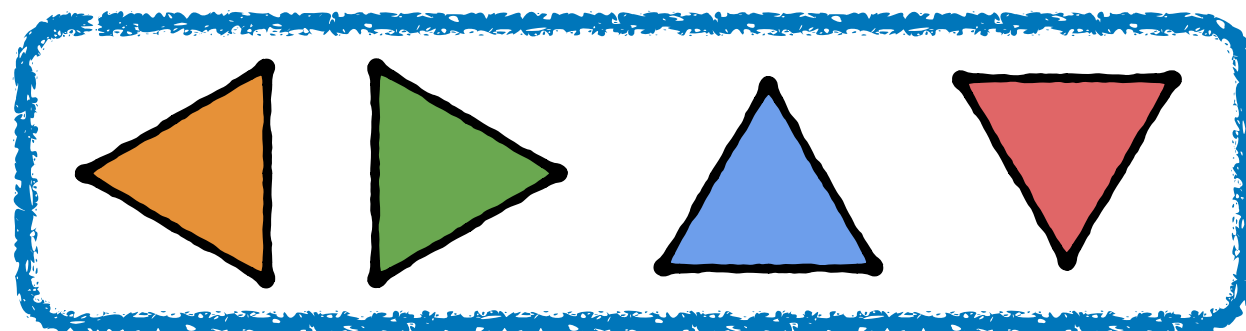
Training



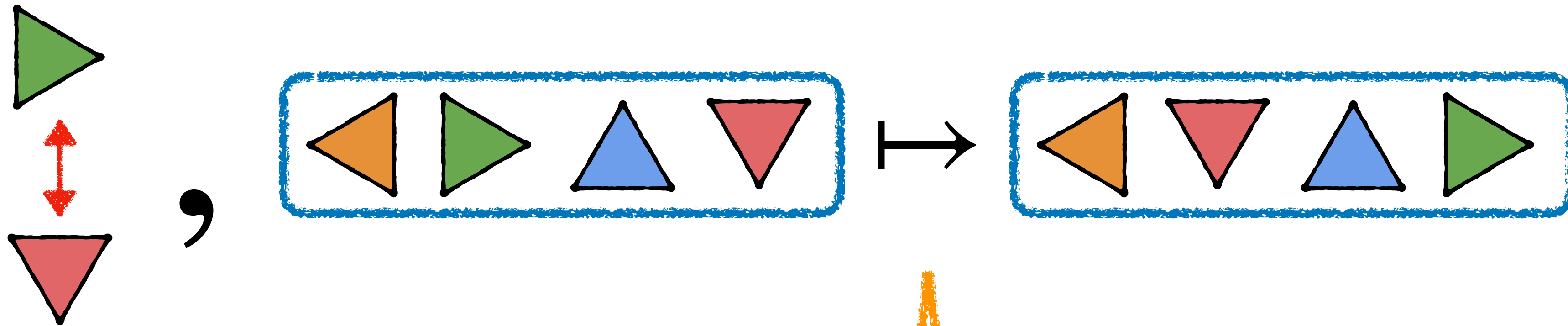
Generalization

Object-replacement: Another View

Example: Replace Object Identity



Object-replacement Operation

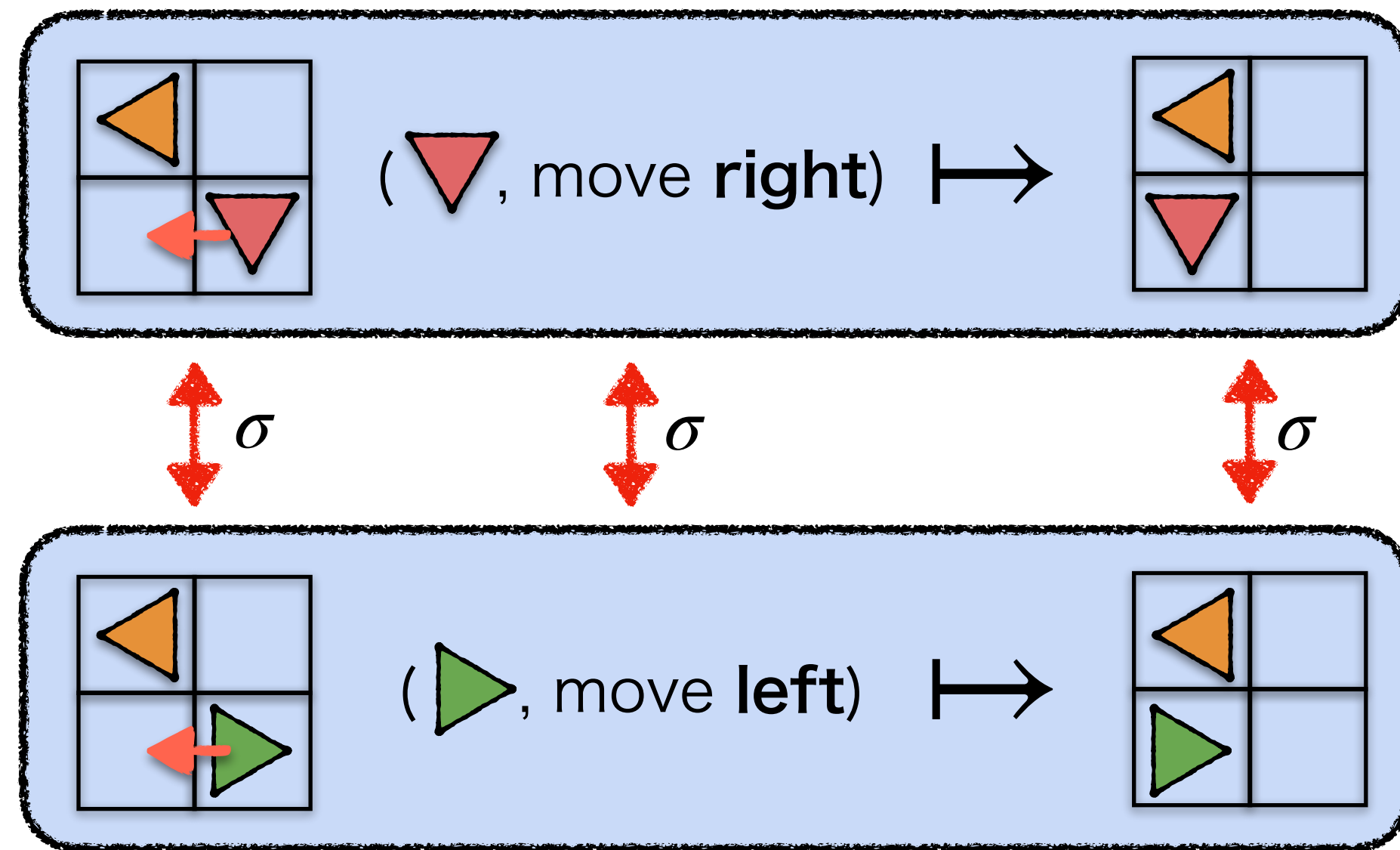


Formally define an operation:

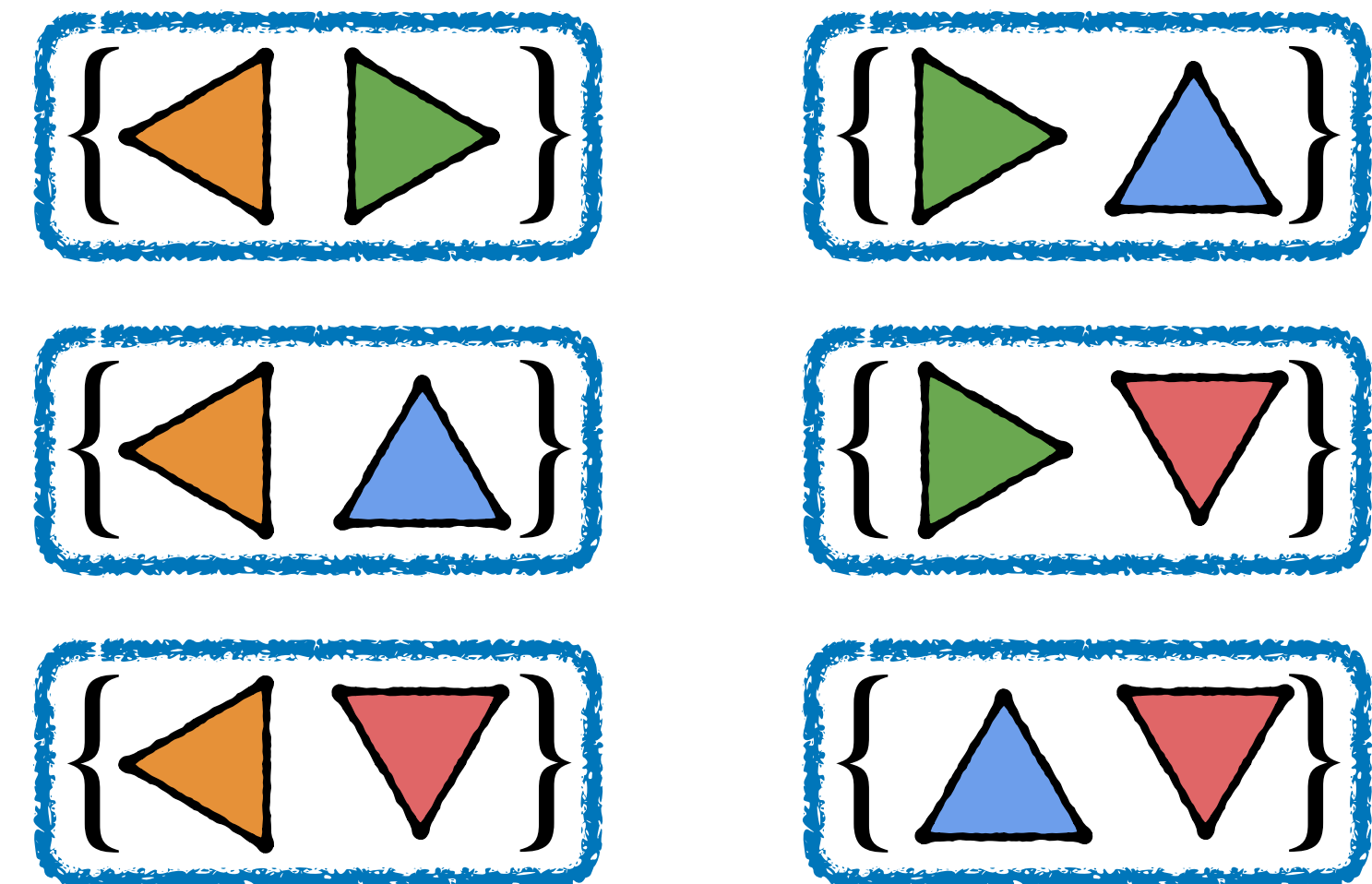
$$p_{\mathbb{L}} : \Sigma_N \times \mathbb{L} \rightarrow \mathbb{L}$$

Size-N Permutation Group

Measure CG with Equivariance Error



Expectation
over
All combinations



How well can the model predict
on *any compositionally different scene*?

$$EE(T_{\perp}) \triangleq \mathbb{E} \left[\left| \hat{T}_{\perp}(s' | s, a) - \hat{T}_{\perp}(\sigma.s' | \sigma.s, \sigma.a) \right| \right]$$

$$\sigma \in \Sigma_N, (s, a, s') \in \mathcal{S}_{\perp} \times \mathcal{A}_{\perp} \times \mathcal{S}_{\perp}$$

Cast **Compositional Generalization**
as **(Permutation) Equivariance Error**
in **Transition Prediction**

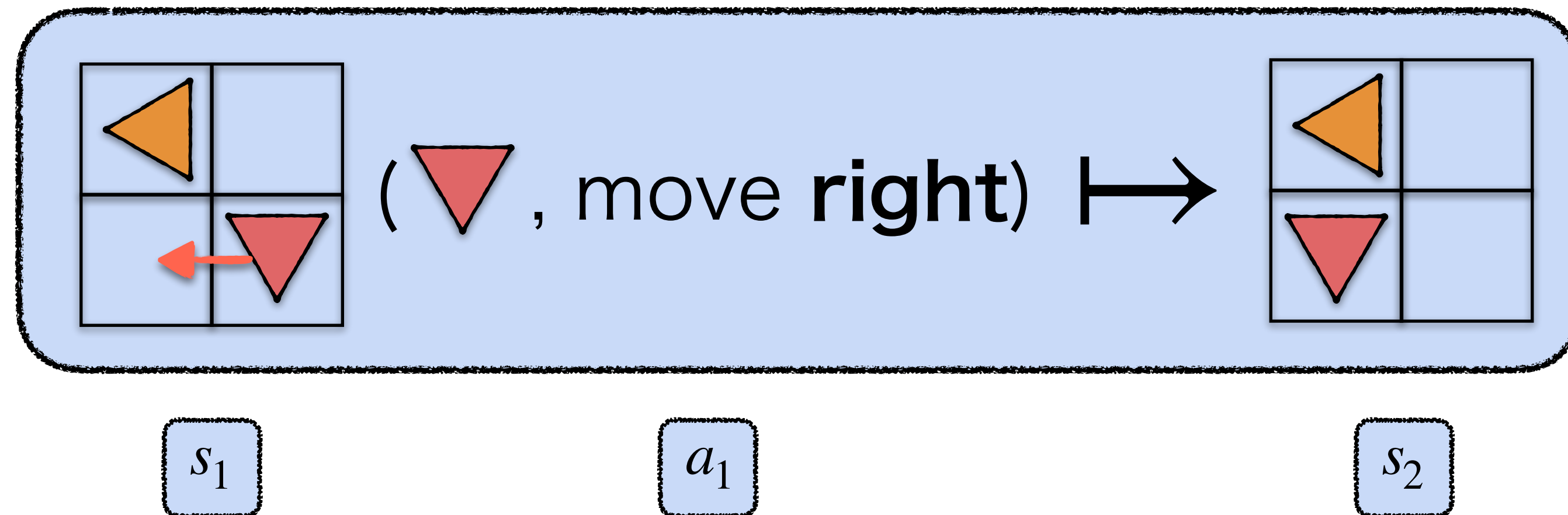
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- **Implementing Compositional Generalization**
- Solving Binding Issue for Compositional Generalization
- Results

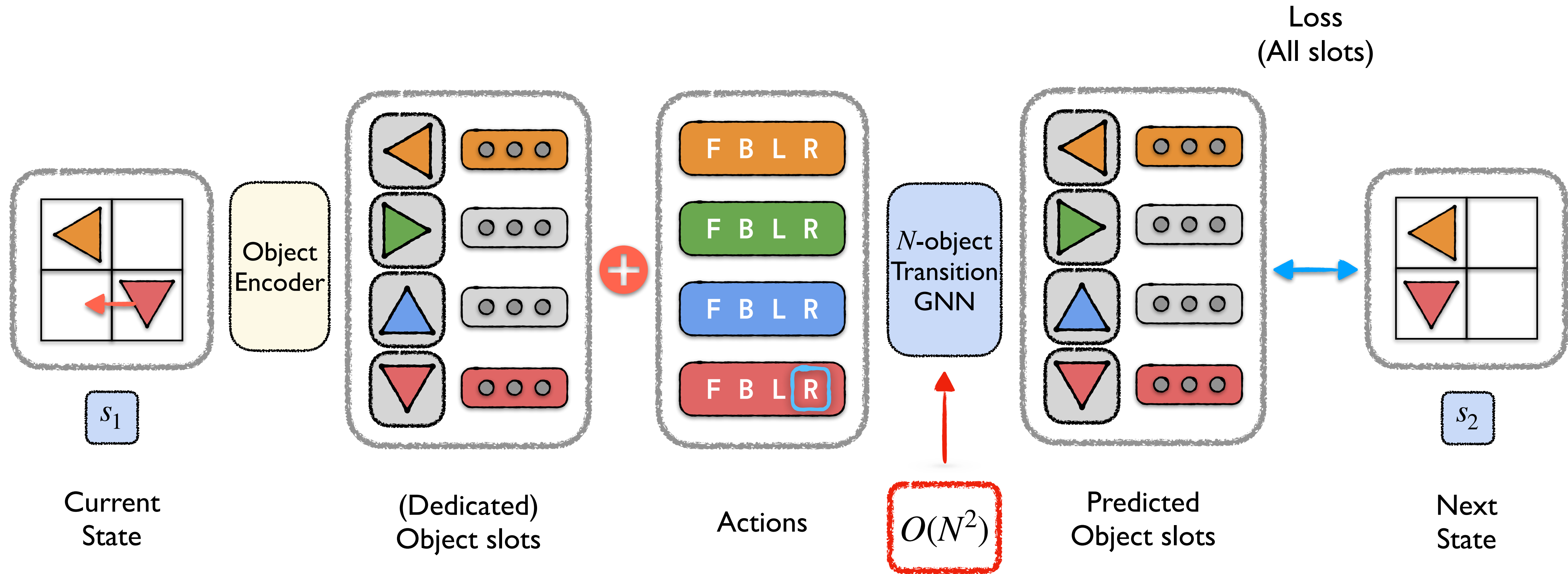
Implementing Compositional Generalization

Recall: World Modeling

Learn a transition model $T : \mathcal{S} \times \mathcal{A} \rightarrow \mathcal{S}$



N-slot GNN: A Naive Solution



Research Questions

- How to **measure** **compositional generalization** (in world modeling)? ✓
- Can we **guarantee** when **compositional generalization** is achieved? ✓
- If possible, how can its **implementation** be **efficient**?

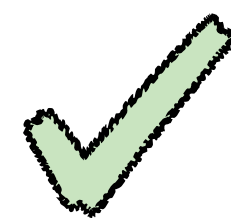
Three Necessary Components

1. **Object** Extraction

2. Actions Concatenated to **Objects**

3. Σ_N -equivariant transition model

N -object GNN
"Library MDP"

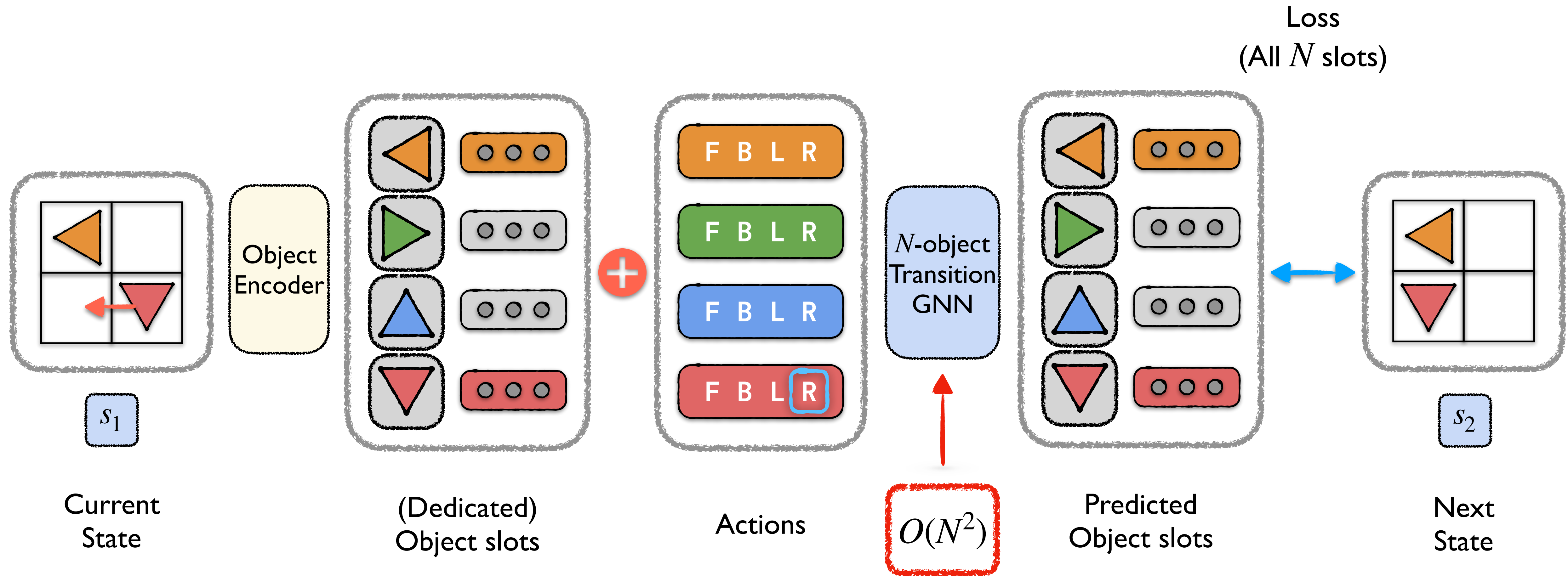


1. **Slot** Extraction

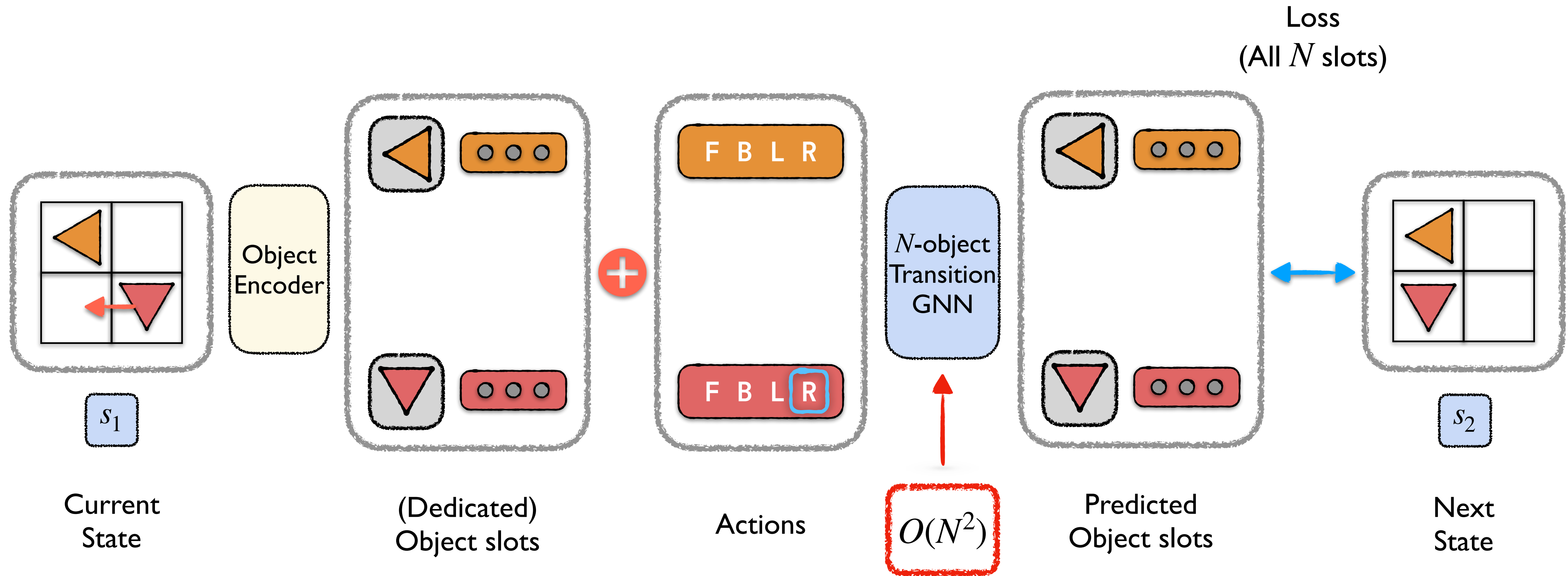
2. Action Binding to **Slots**

3. Σ_K -equivariant latent transition model

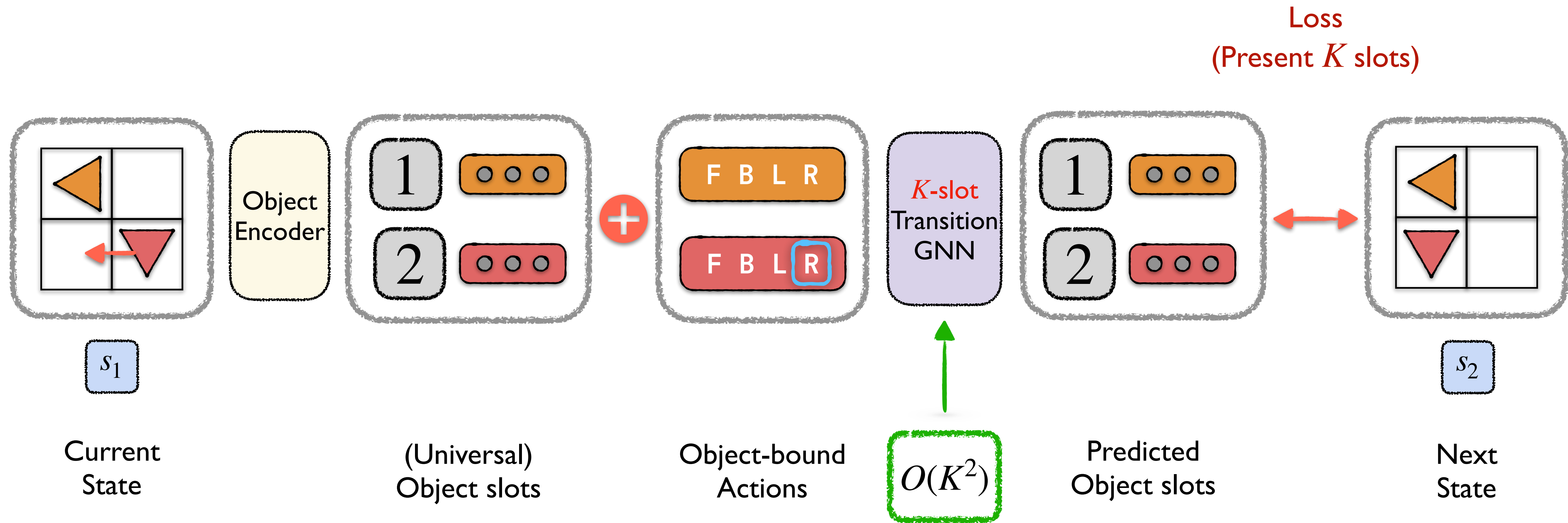
K-slot GNN: More Efficient?



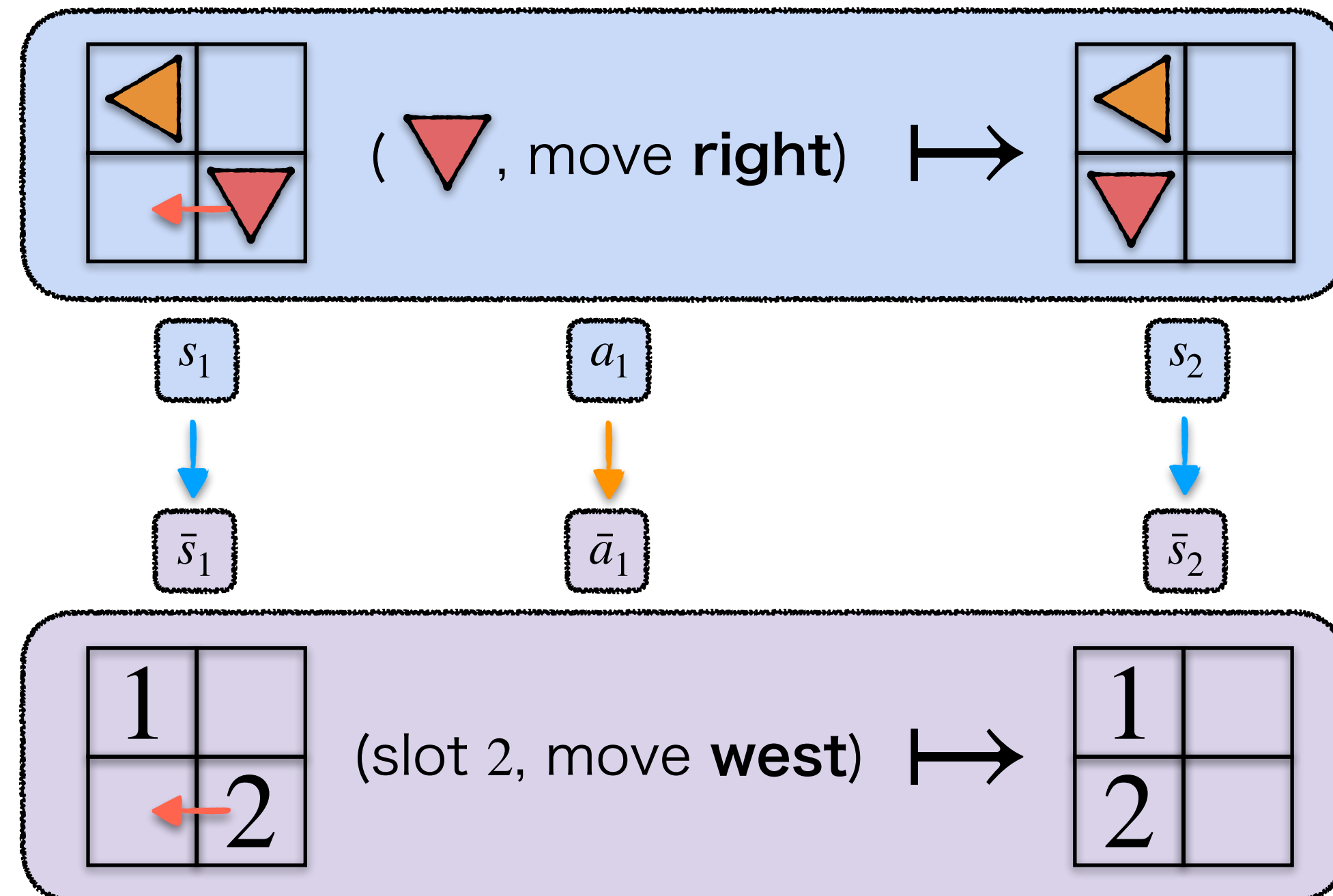
K-slot GNN: More Efficient?



K-slot GNN: *More Efficient?*



Slot MDP



Ideally: want a **canonical MDP** such that **all scenes** are **isomorphic** to it

“Slot MDP”
(Any slot can bind any object & action)

Challenge: *no canonical order* of objects or slots

Motivating Visualization

Challenge: *no canonical order* of objects or slots

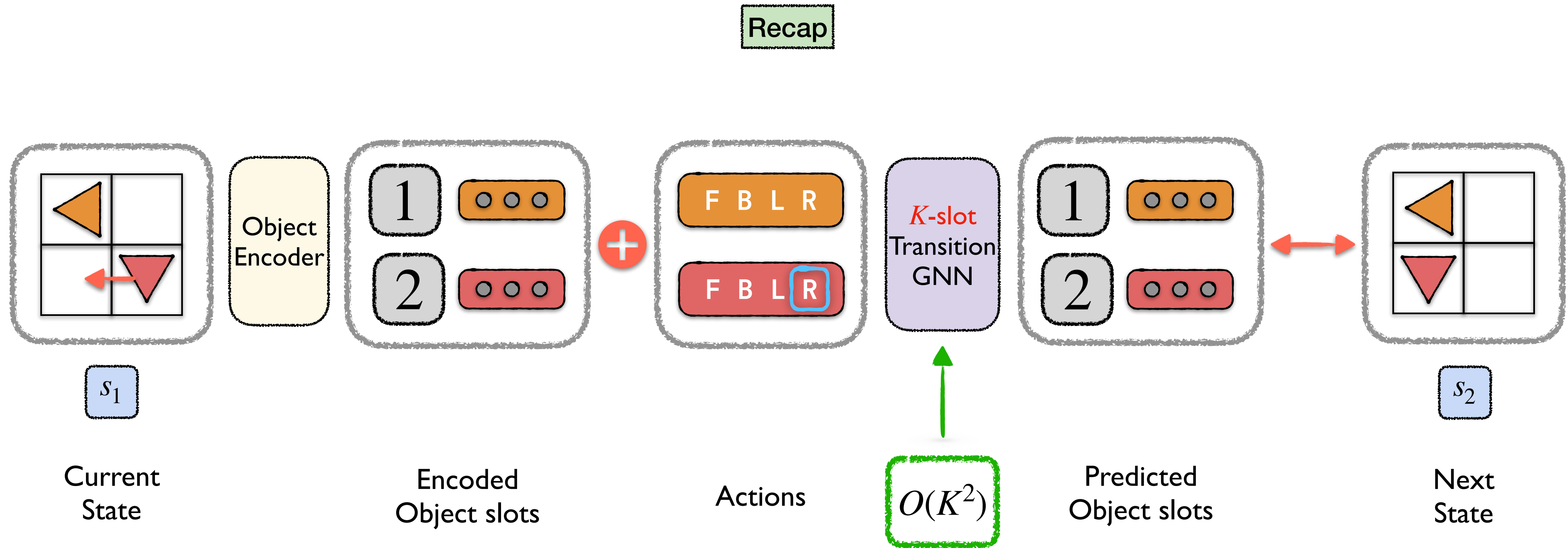


Outline

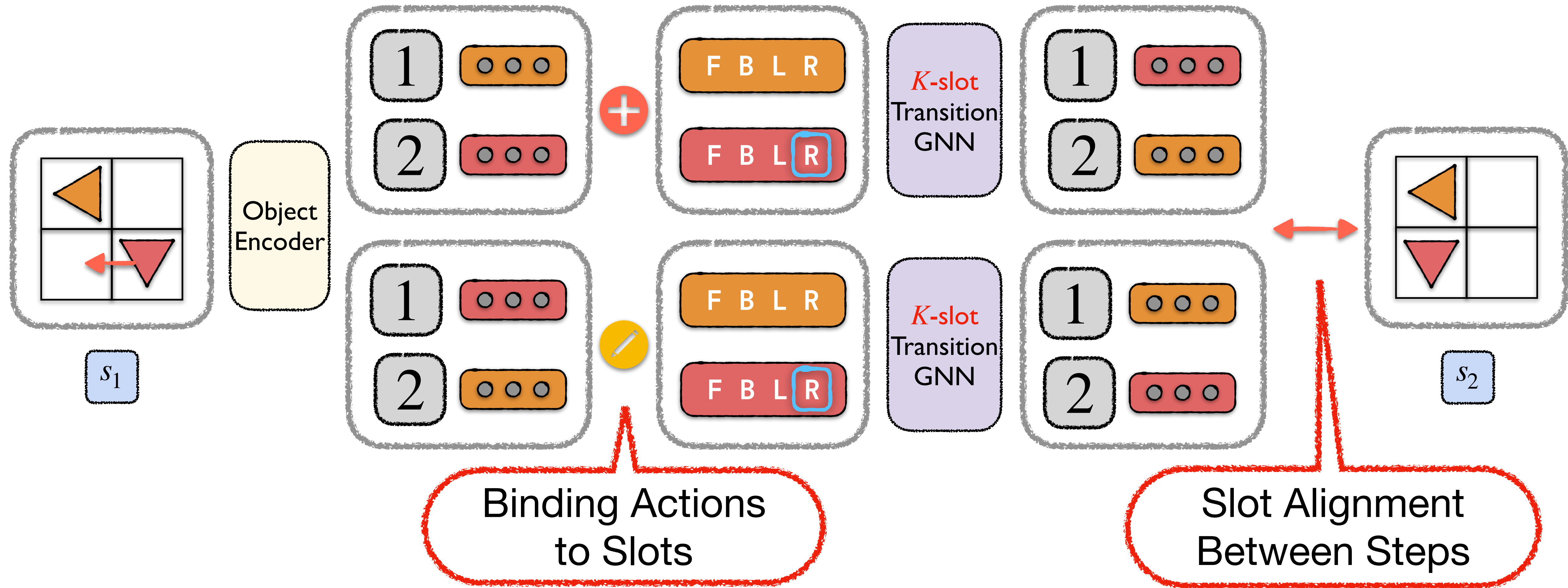
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Solving Binding Issue for Compositional Generalization

K-slot GNN: *Binding Issue*

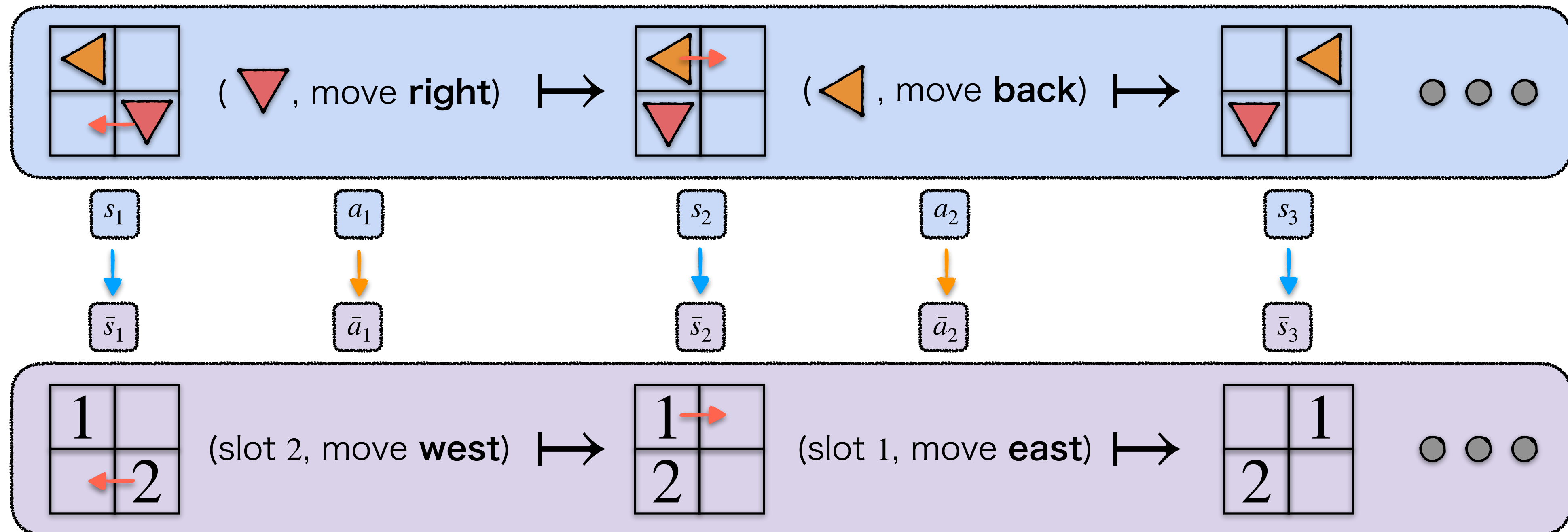


K-slot GNN: *Binding Issue*

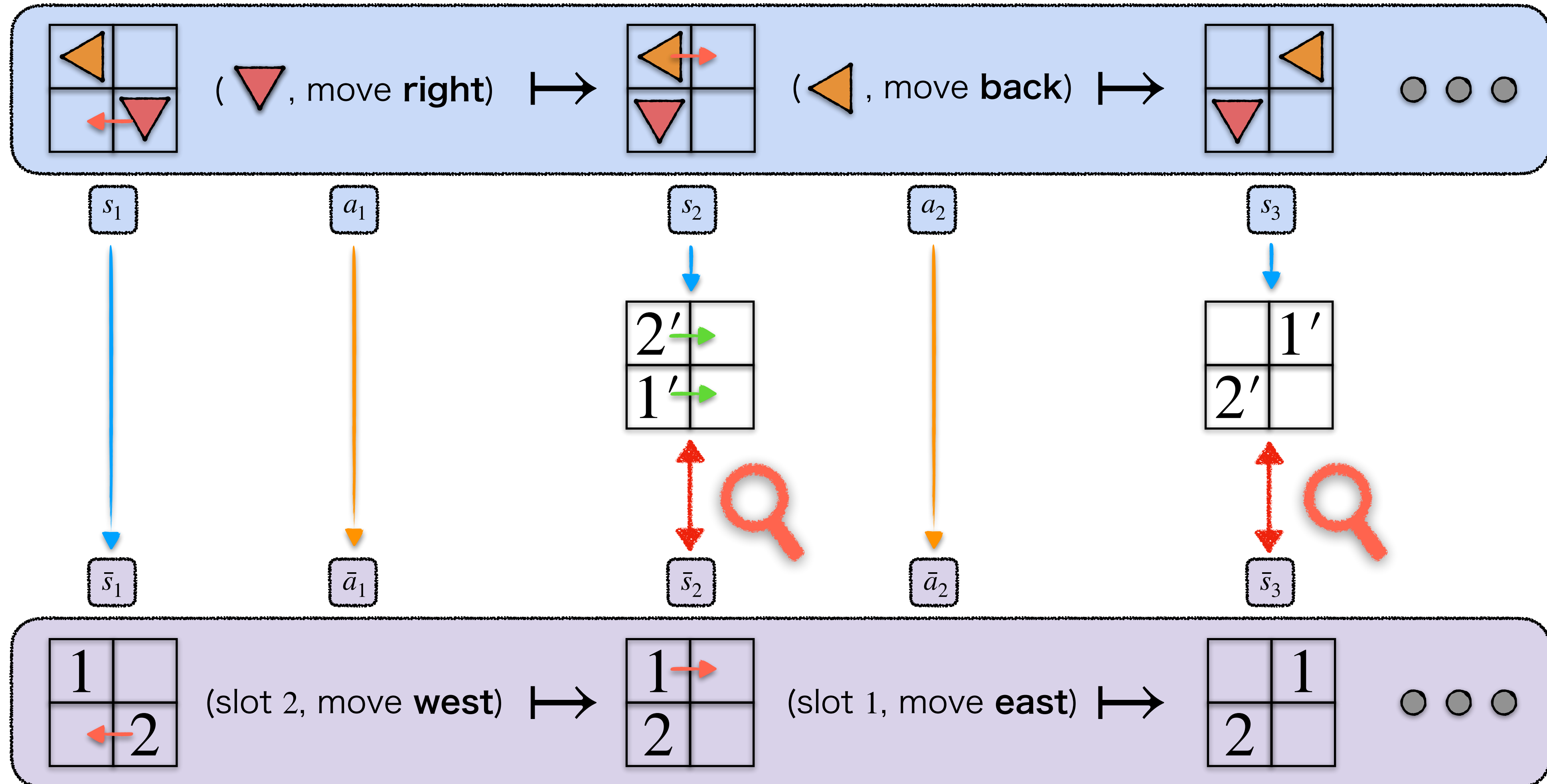


Binding Issue in Slot MDP

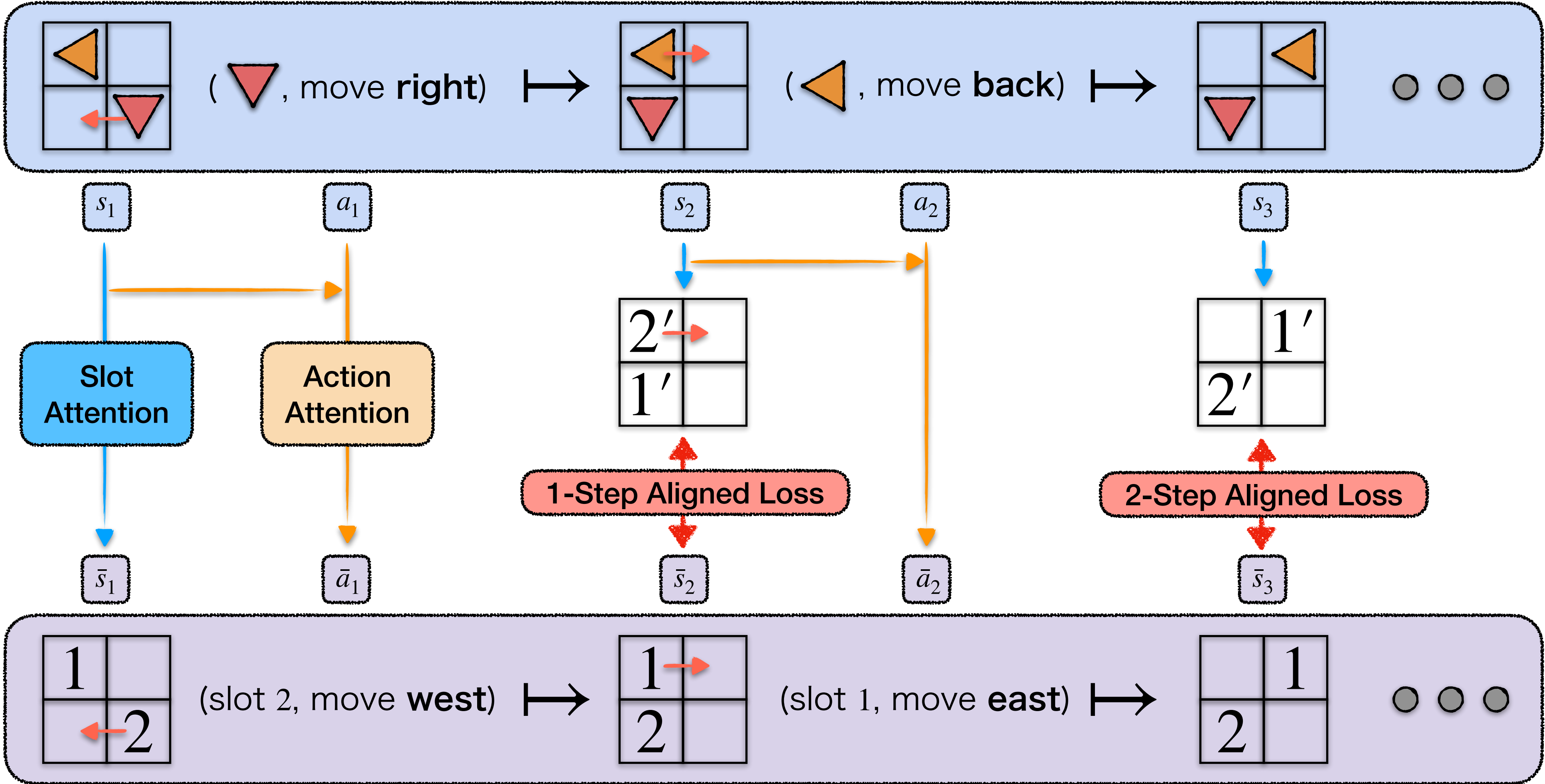
Another View: Multi-step World Model Inference



Binding Issue in Slot MDP



Method: Action Attention & Aligned Loss



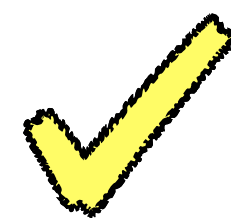
Three Necessary Components

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3. Σ_N -equivariant transition model

N -object GNN
"Library MDP"



1. **Slot** Extraction

2. Action Binding to **Slots**

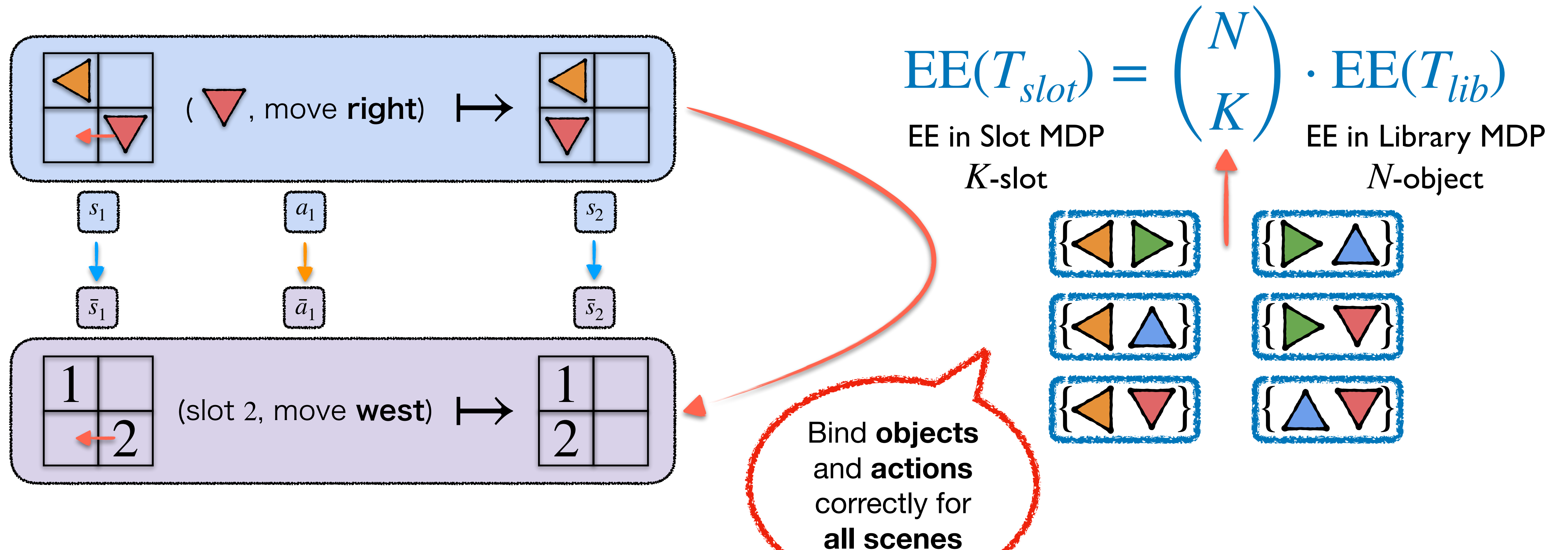
3. Σ_K -equivariant latent transition model

K -slot GNN
"Slot MDP"



Key Theorem: EE in Slot MDP

Theorem (informal): If **object binding** and **action binding** are correct, the equivariance error is related by:



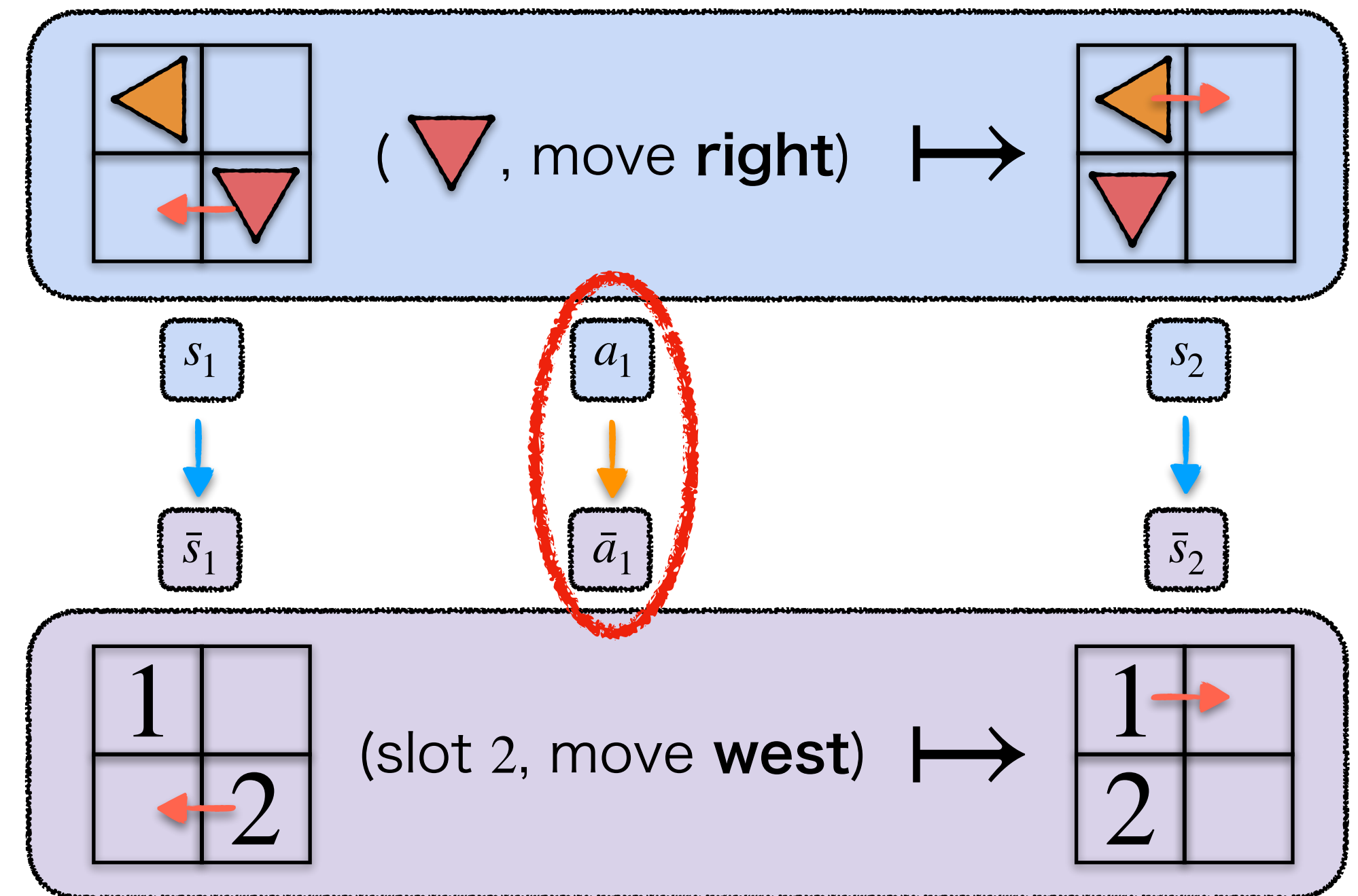
Corollary: CG in Latent Space

$$\underbrace{0}_{\text{EE in Slot MDP}} = \underbrace{\begin{pmatrix} N \\ K \end{pmatrix}}_{\text{EE in Library MDP}} \cdot \text{EE}(T_{lib})$$

K -slot N -object

If object binding and action binding are correct:

$$\bar{\sigma} \cdot \phi(s) = \phi(\sigma \cdot s) \quad \text{and} \quad \bar{\sigma} \cdot \alpha_s(a) = \alpha_{\sigma \cdot s}(\sigma \cdot a)$$

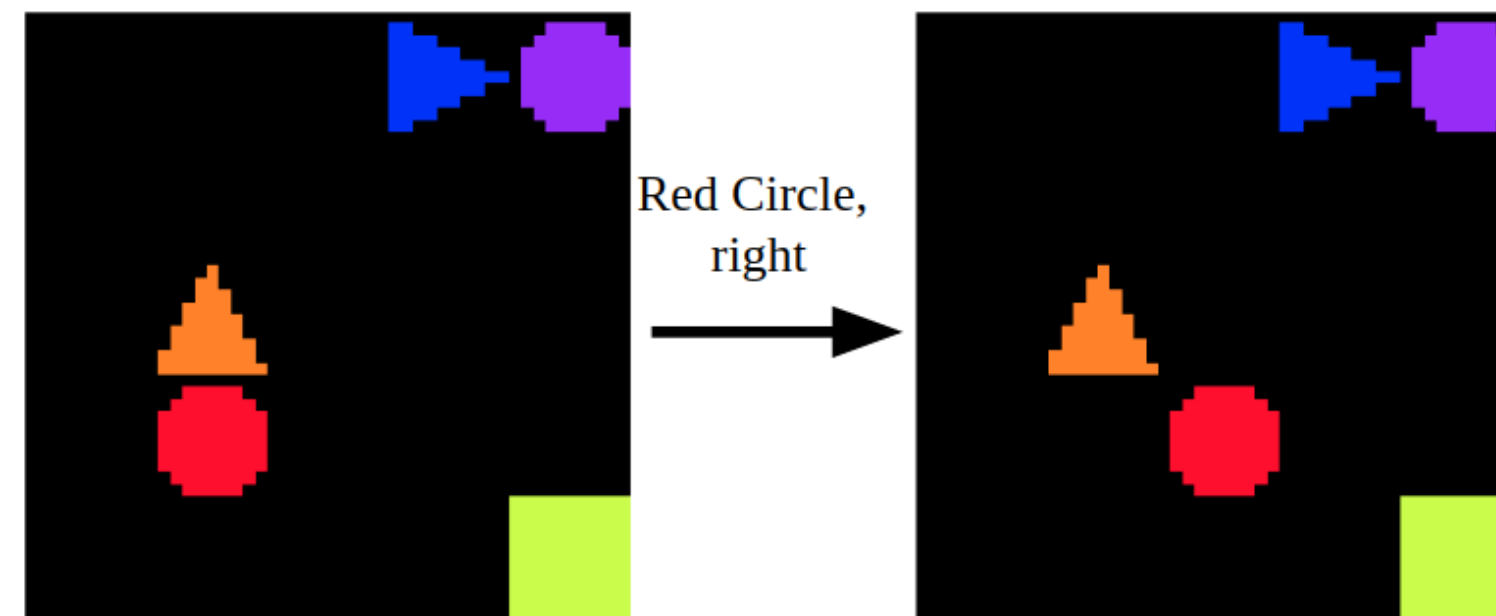


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Environment (Block Pushing)



Object Library: Block Pushing

Random objects (locations, color, shapes)

$K = 5, N = 5, 10, 20, 30$

Experimental Setup

- **Exact CG:** Σ_N -equivariant WM
- **No CG:** break each component
- **Soft CG:** Learned Σ_K -equivariant WM



1. **Slot** Extraction

2. Action Binding to **Slots**

3. Σ_K -equivariant latent transition model

Measuring CG in Practice

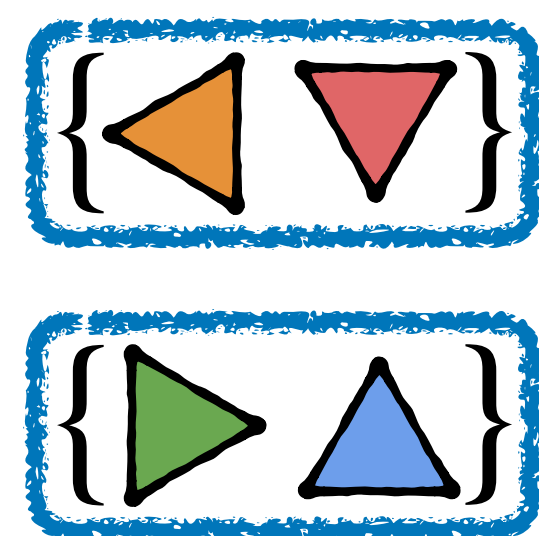
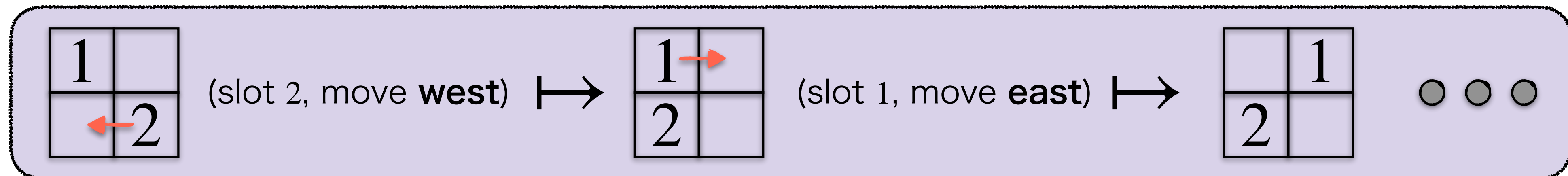
$$EE(T_{lib}) \triangleq \mathbb{E} \left[\left| \hat{T}_{\perp}(s' | s, a) - \hat{T}_{\perp}(\sigma \cdot s' | \sigma \cdot s, \sigma \cdot a) \right| \right]$$

$\sigma \in \Sigma_N, (s, a, s') \in \mathcal{S}_{\perp} \times \mathcal{A}_{\perp} \times \mathcal{S}_{\perp}$

$$EE(T_{slot}) = \binom{N}{K} \cdot EE(T_{lib})$$

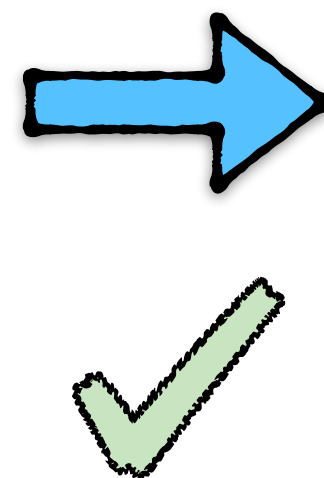
✗

Binding



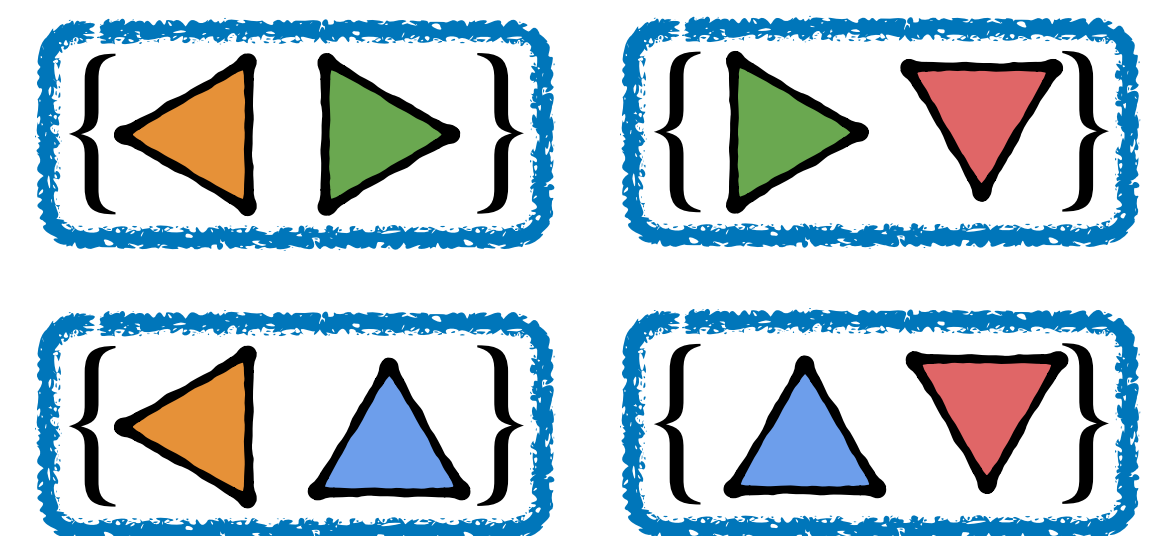
Training

“Training Sentences”



Generalization

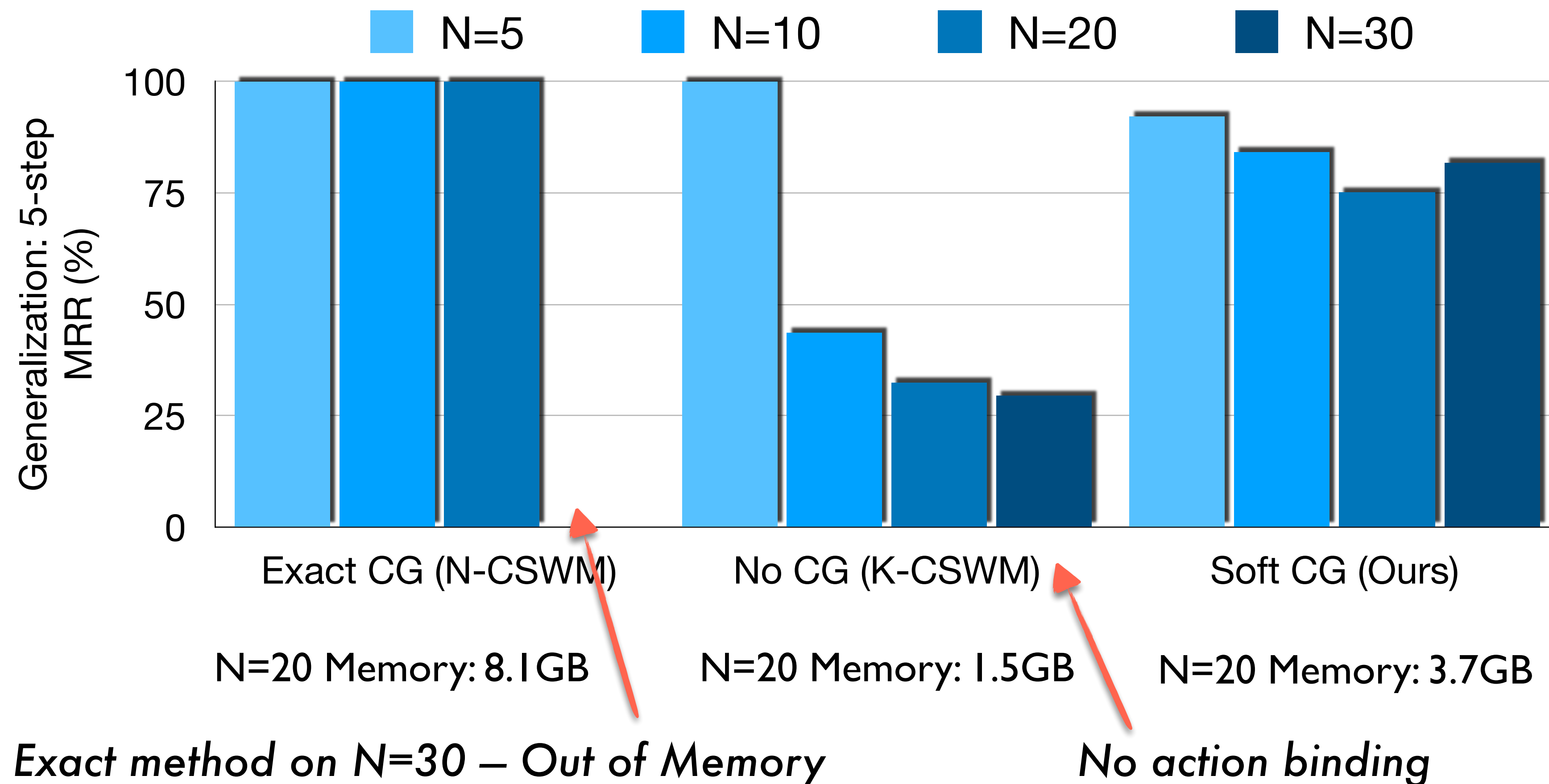
“Test Sentences”



[Keysers et al., ICLR 2020, “Measuring Compositional Generalization: A Comprehensive Method on Realistic Data”]

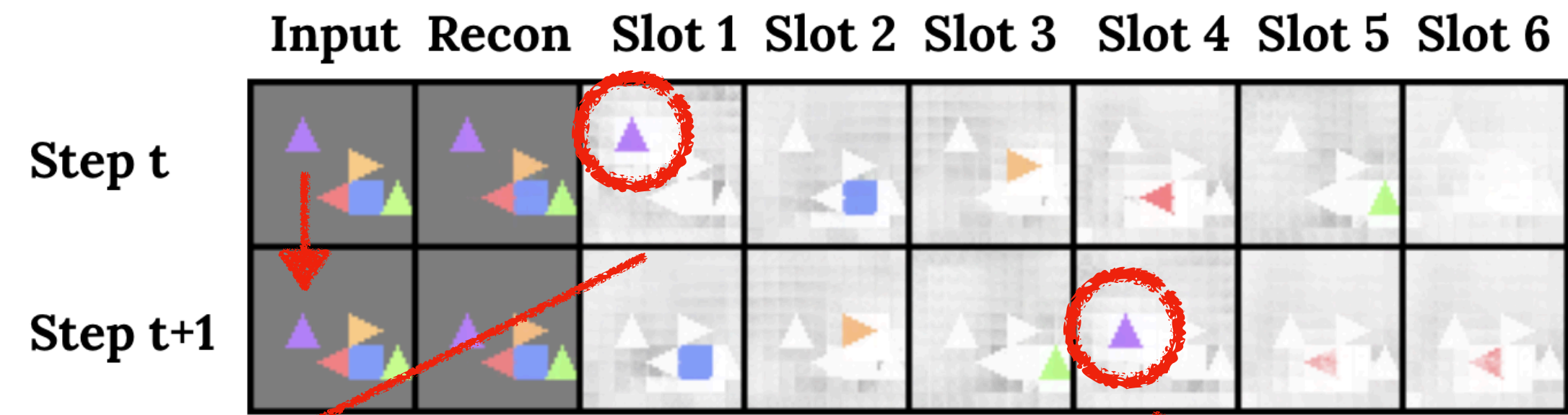
Quantitative Results

5-step MRR (%) on Novel Test Scenes — Higher is better



[Kipf et al., ICLR 2020, Contrastive Learning of Structured World Models]

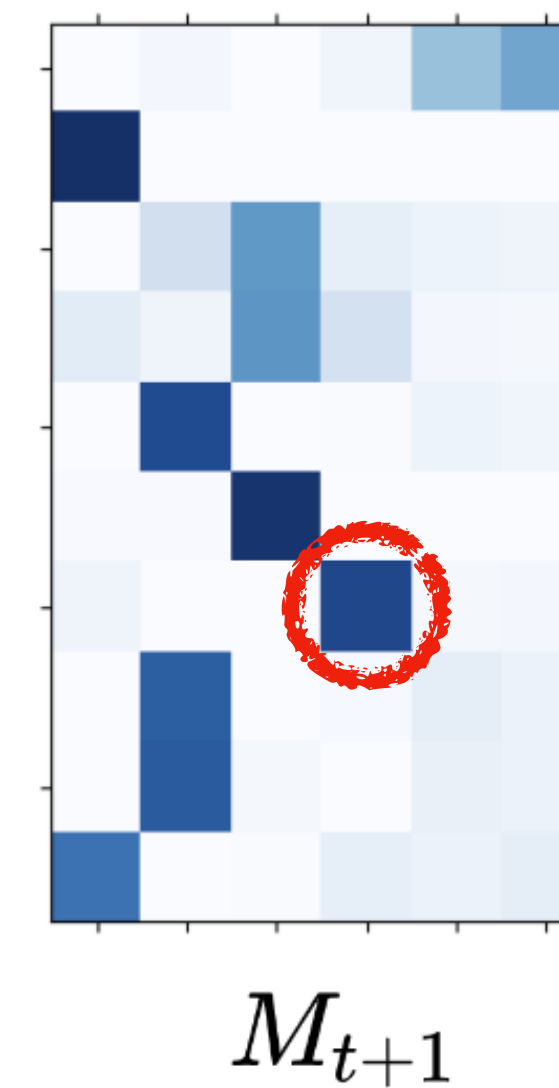
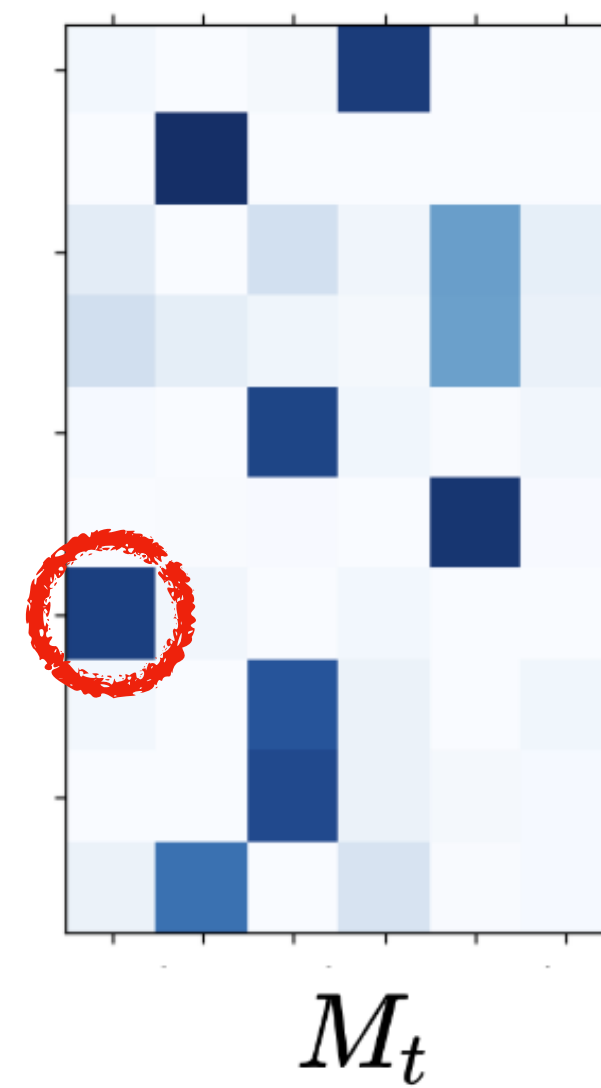
Binding Visualization



5+1 rows:
K=5 slots + 1 background

10 columns:
N=10 objects

Object 7



Found object identity
through actions
(Objects unknown)

Summary

Study compositional generalization in object-oriented world modeling

- Introduce a conceptual environment (Object Library)
- Formalize compositional generalization (by equivariance error)
- Propose a soft approach with guarantees
- Practically measure compositional generalization

More Information

Check out our project website:



<http://lfzhao.com/oowm/>
zhao.linf@northeastern.edu