

Rodrigues Network for Learning Robot Actions

Jialiang Zhang^{1,2,*}, Haoran Geng^{3,*}, Yang You^{2,*}, Congyue Deng^{1,2},
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ICLR
International Conference On
Learning Representations

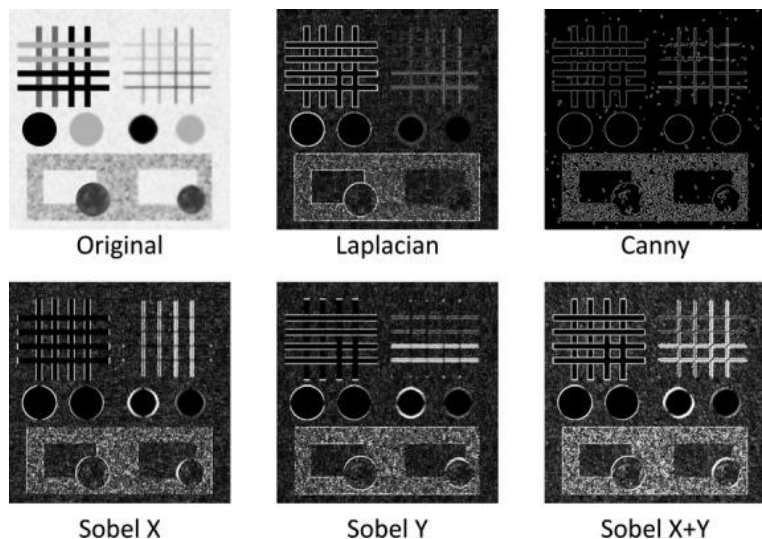
Motivation



How to Embed Inductive Bias into Networks: CNN

2D Image Filtering

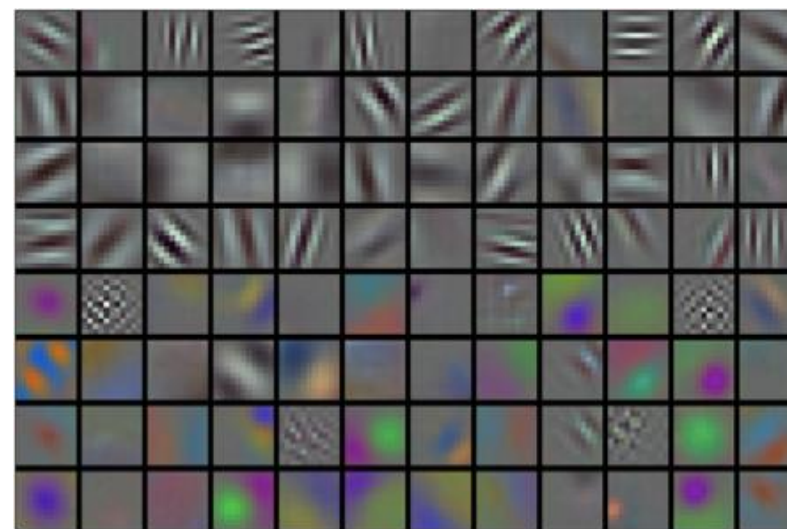
Fixed hand-crafted filters



DETECT features

2D Image Convolution

Learnable neural kernels



LEARN features

Uchida et. al. Image processing and recognition for biological images.
Krizhevsky et. al. ImageNet classification with deep convolutional neural networks.

How to Embed Inductive Bias into Networks: CNN

2D Image Filtering

Fixed hand-crafted filters

2D Image Convolution

Learnable neural kernels

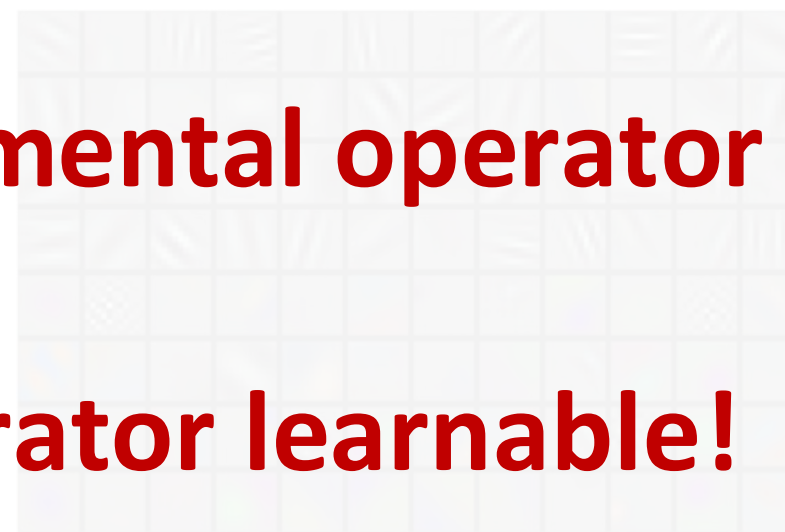
Step 1: Understand structure of modality



Step 2: Identify fundamental operator



Step 3: Make that operator learnable!



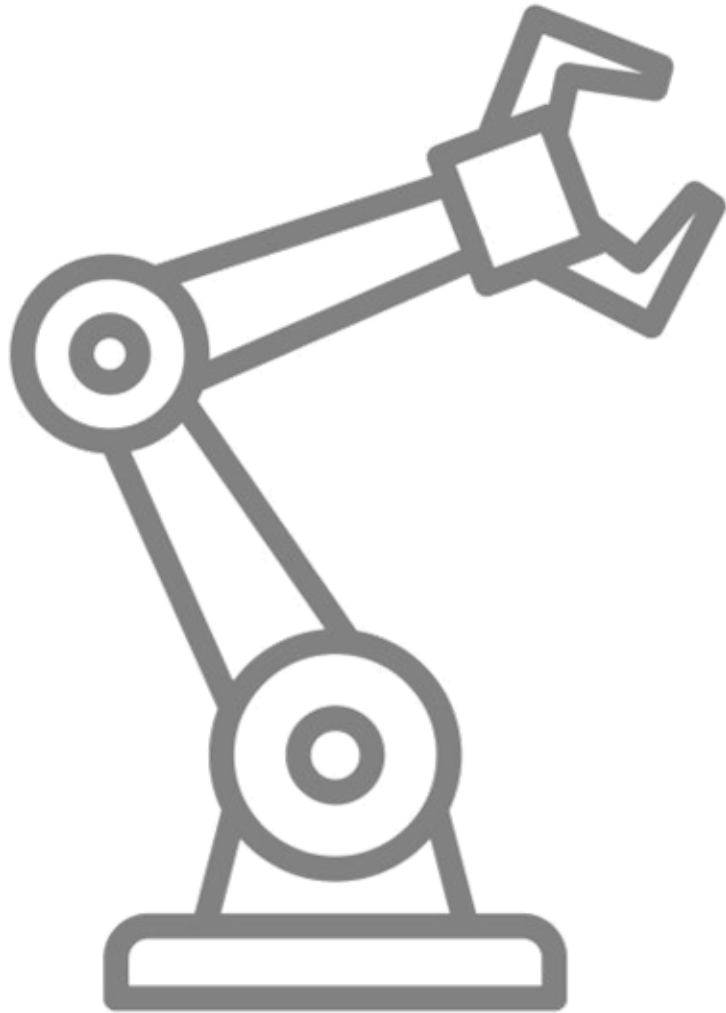
DETECT features

LEARN features

Uchida et. al. Image processing and recognition for biological images.

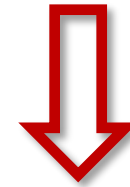
Krizhevsky et. al. ImageNet classification with deep convolutional neural networks.

Core Structure of Actions: Articulated Kinematics



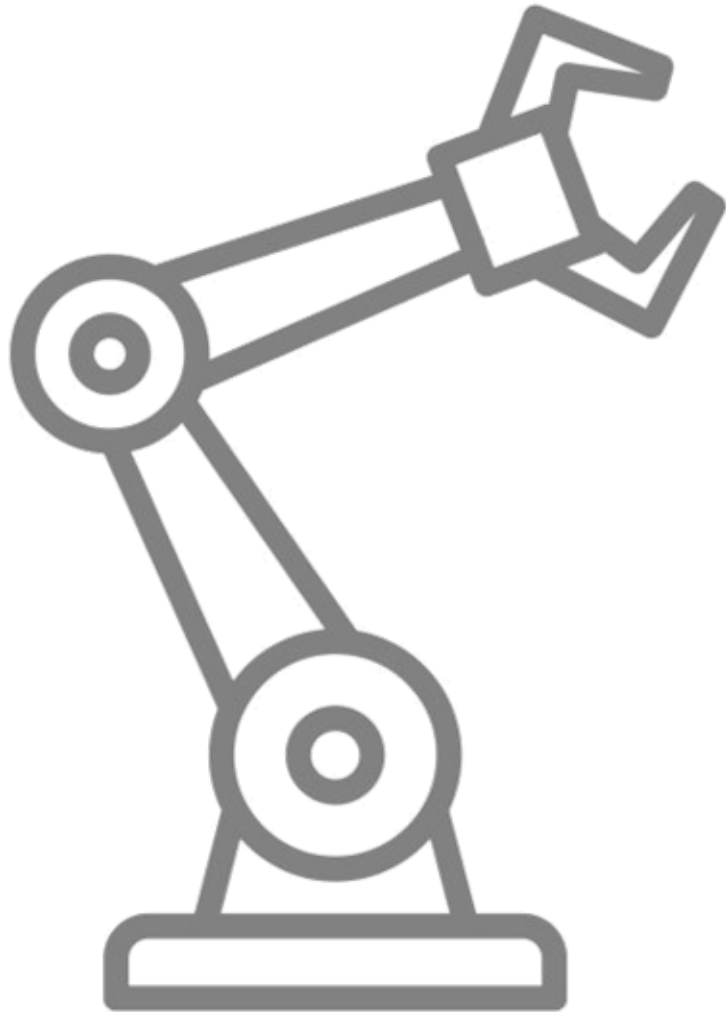
Action: θ

**Articulated
Kinematics**



3D world

Key Operator: Rodrigues' Rotation Formula



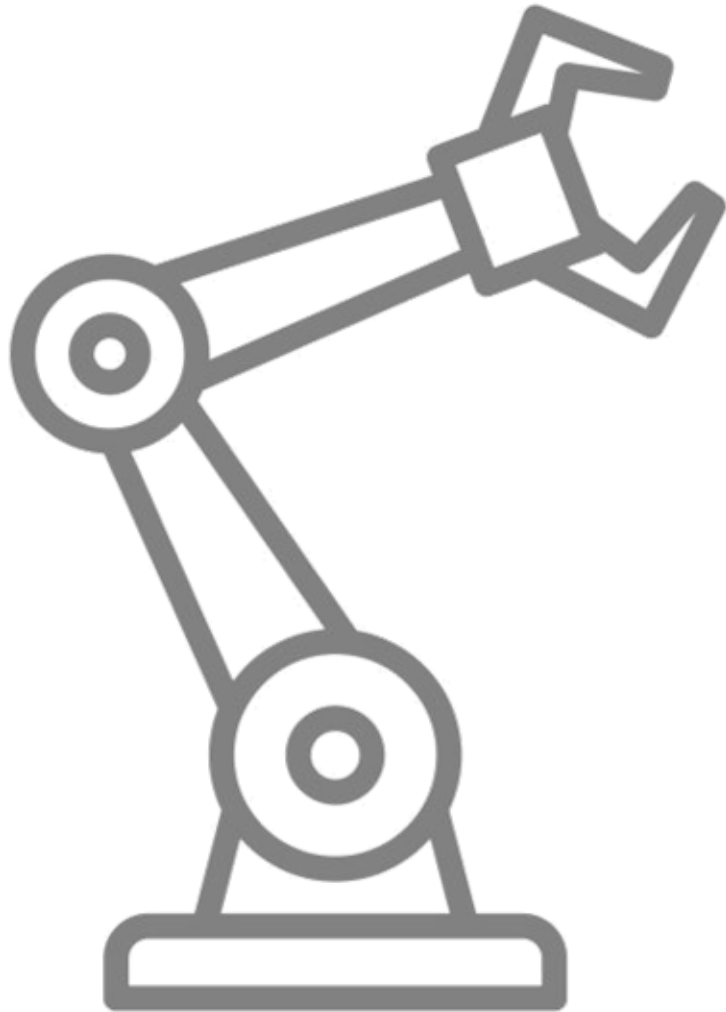
**Articulated
Kinematics**



**Stack of
Rodrigues'
Rotation
Formula**

Details omitted
(see paper)

Key Operator: Rodrigues' Rotation Formula



**Articulated
Kinematics**



**Stack of
Rodrigues'
Rotation
Formula**



**Inductive bias
of Network?**



Rodrigues' Rotation Formula

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$

Rodrigues' Rotation Formula

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$



Rodrigues' Rotation Formula

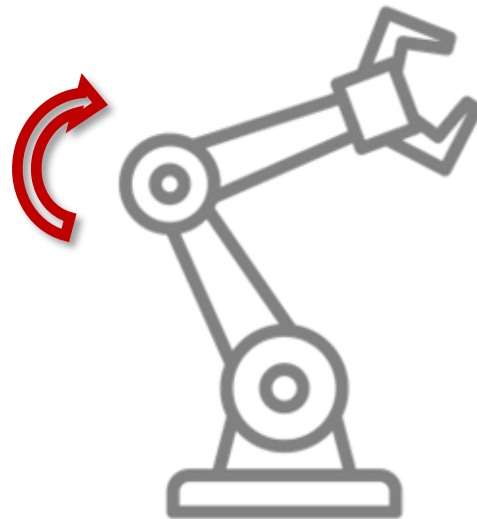
$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$

$$\begin{bmatrix} 0 & -\omega_z & \omega_y \\ \omega_z & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{bmatrix}$$



Rodrigues' Rotation Formula

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$



Rodrigues' Rotation Formula

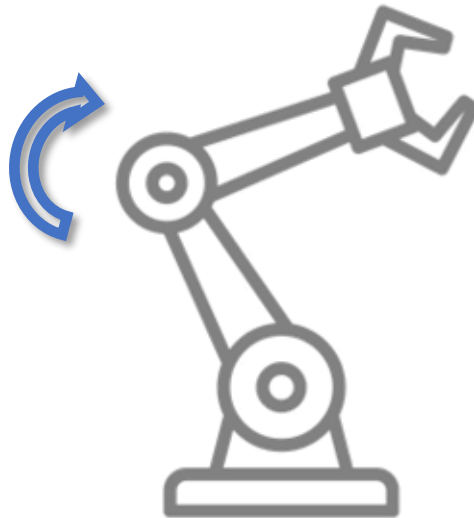
Joint Space



World Space

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$

θ



R

Rodrigues' Rotation Formula

Action-independent

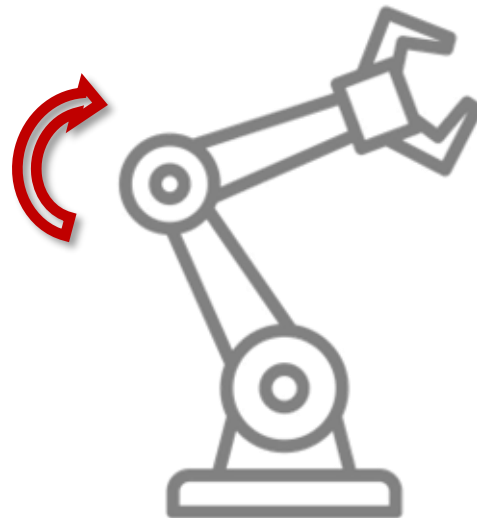
$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$



Rodrigues' Rotation Formula

Action-dependent

$$\mathbf{I}_3 + [\hat{\omega}] \sin \theta + [\hat{\omega}]^2 (1 - \cos \theta)$$



Essence of Rodrigues' Rotation Formula

Action-dependent

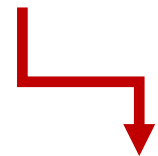
$$\mathbf{I}_3 + [\hat{\omega}] \sin \theta + [\hat{\omega}]^2 (1 - \cos \theta)$$

Re-order

Essence of Rodrigues' Rotation Formula

Action-dependent

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$



Re-order

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Essence of Rodrigues' Rotation Formula

Action-dependent

$$\mathbf{I}_3 + [\hat{\boldsymbol{\omega}}] \sin \theta + [\hat{\boldsymbol{\omega}}]^2 (1 - \cos \theta)$$

 Re-order

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Essence of Rodrigues' Rotation Formula

Action-independent

$$\mathbf{I}_3 + [\hat{\omega}] \sin \theta + [\hat{\omega}]^2 (1 - \cos \theta)$$



Re-order

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Essence of Rodrigues' Rotation Formula

Factorize action-dependent/independent

$$\mathbf{I}_3 + [\hat{\omega}] \sin \theta + [\hat{\omega}]^2 (1 - \cos \theta)$$



$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Essence of Rodrigues' Rotation Formula

Rodrigues' Rotation Formula

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

sin/cos
Joint
Space

$$\begin{bmatrix} 1 \\ \sin \theta \\ \cos \theta \end{bmatrix}$$

Linear



R

World
Space

Make Rodrigues' Rotation Formula Learnable

Rodrigues' Rotation Formula

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Make Rodrigues' Rotation Formula Learnable

Rodrigues' Rotation Formula

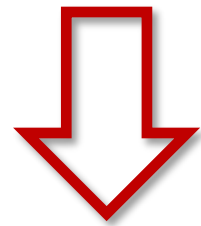
$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$

Fixed parameters

Make Rodrigues' Rotation Formula Learnable

Rodrigues' Rotation Formula

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$



Fixed parameters

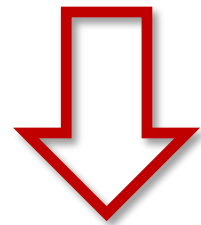
Learnable kernels

$$\mathbf{W}_{\text{bias}} + \mathbf{W}_{\text{sin}} \sin \theta + \mathbf{W}_{\text{cos}} \cos \theta$$

Make Rodrigues' Rotation Formula Learnable

Rodrigues' Rotation Formula

$$\mathbf{A} + \mathbf{B} \sin \theta + \mathbf{C} \cos \theta$$



Fixed parameters

Neural Rodrigues Operator

Learnable kernels

$$\mathbf{W}_{\text{bias}} + \mathbf{W}_{\text{sin}} \sin \theta + \mathbf{W}_{\text{cos}} \cos \theta$$

Make Rodrigues' Rotation Formula Learnable

Step 1 Structure: Joint Space \leftrightarrow World Space

Step 2 Operator: Rodrigues' Rotation Formula

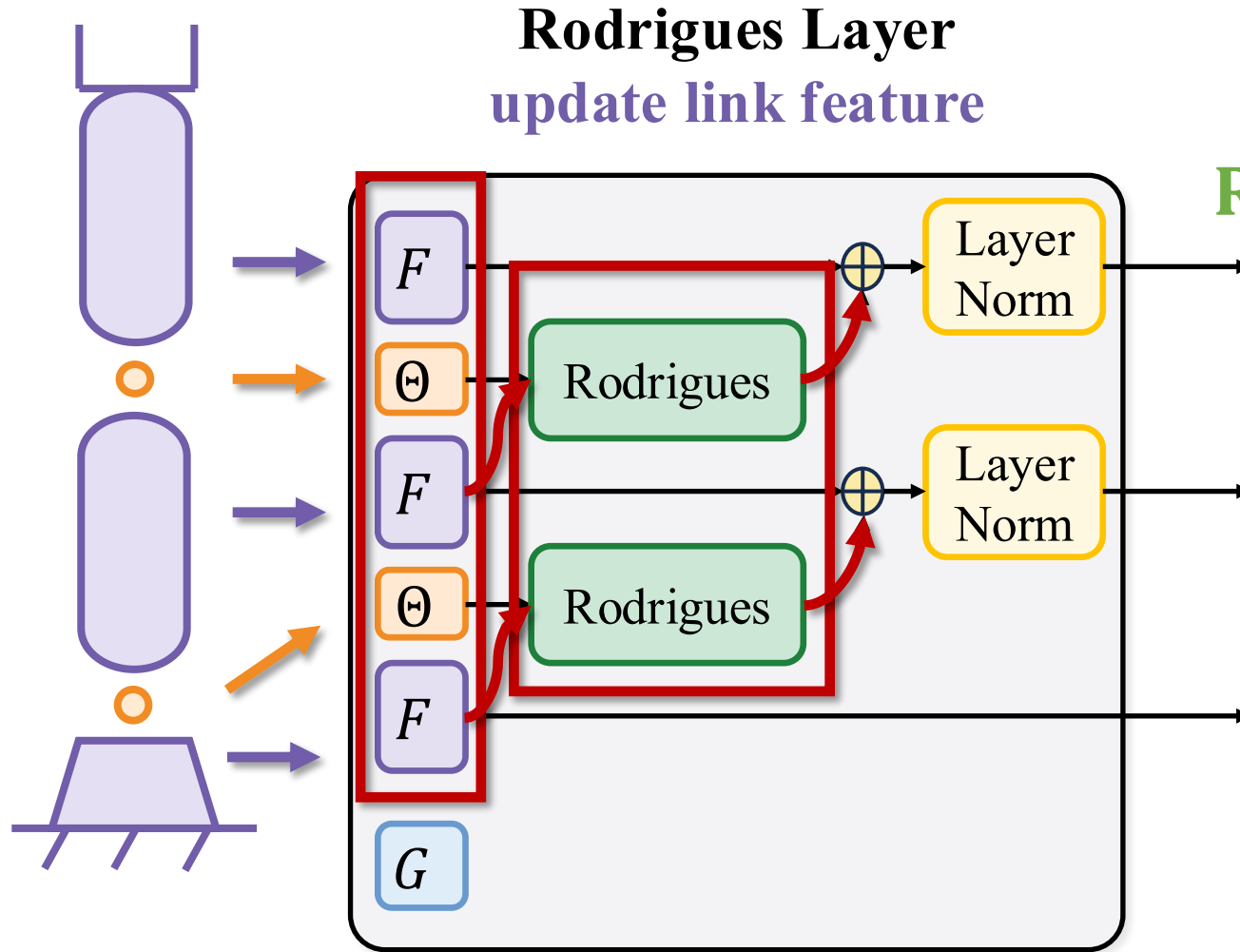
Step 3 Learnable: Neural Rodrigues Operator

Articulated kinematics \Rightarrow **Learnable Generalization**

(Recall: CNN)

(see paper)

Build Rodrigues Network



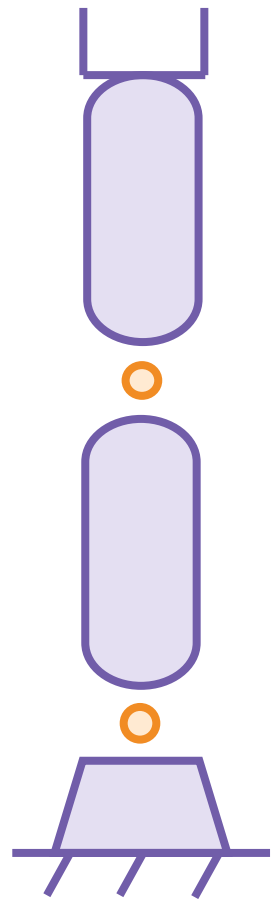
$$\mathbf{R} = \mathbf{W}_{\text{bias}} + \mathbf{W}_{\text{sin}} \sin \theta + \mathbf{W}_{\text{cos}} \cos \theta$$

$$F_{\text{child}} = \mathbf{R} F_{\text{parent}}$$

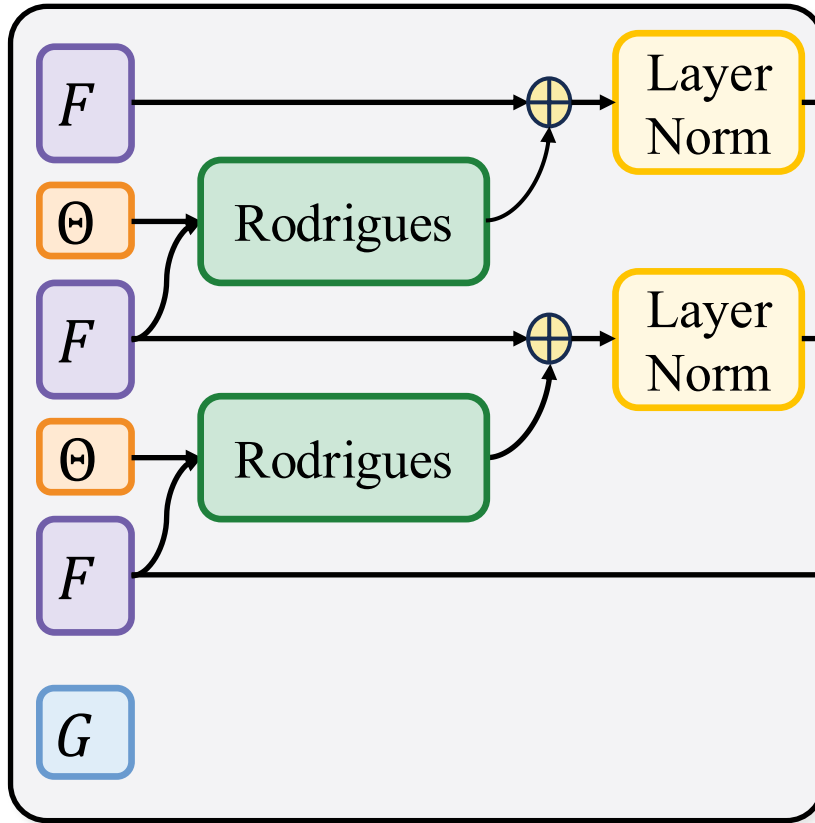
Intuitions omitted
(see paper)

**Hierarchical feature
passing pattern**

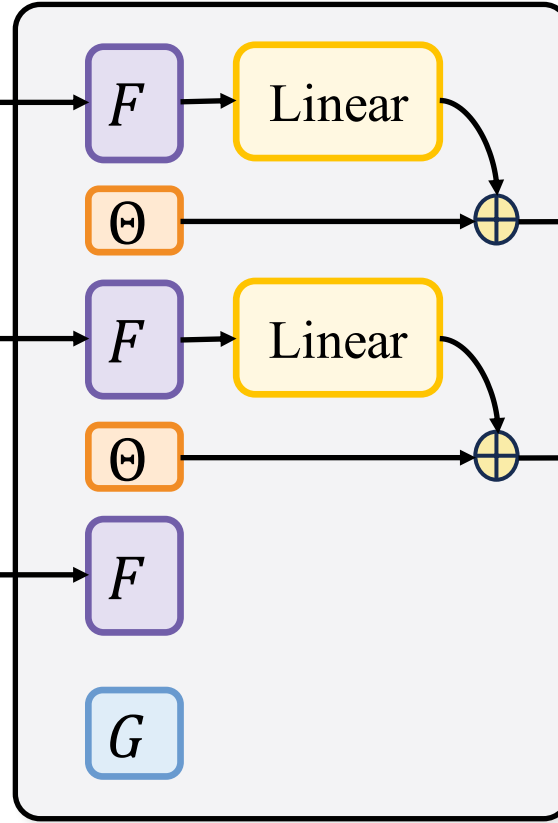
Build Rodrigues Network



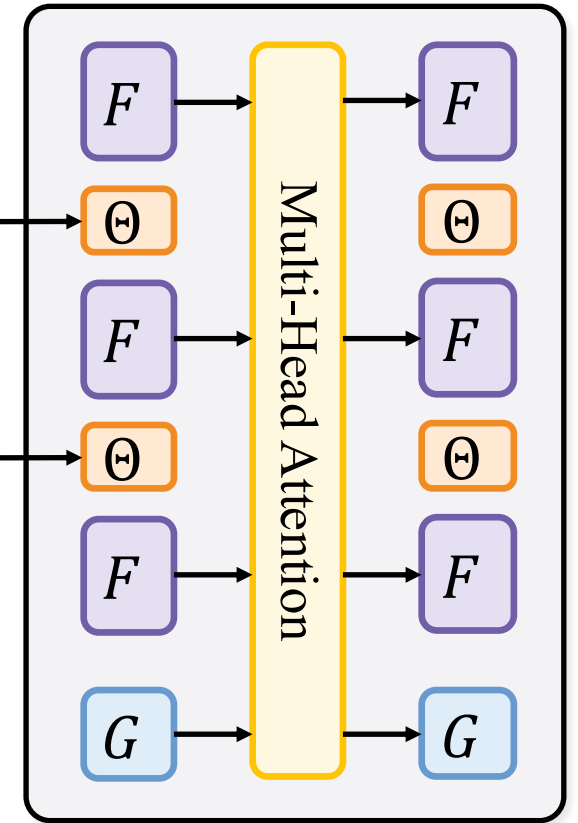
Rodrigues Layer
update link feature



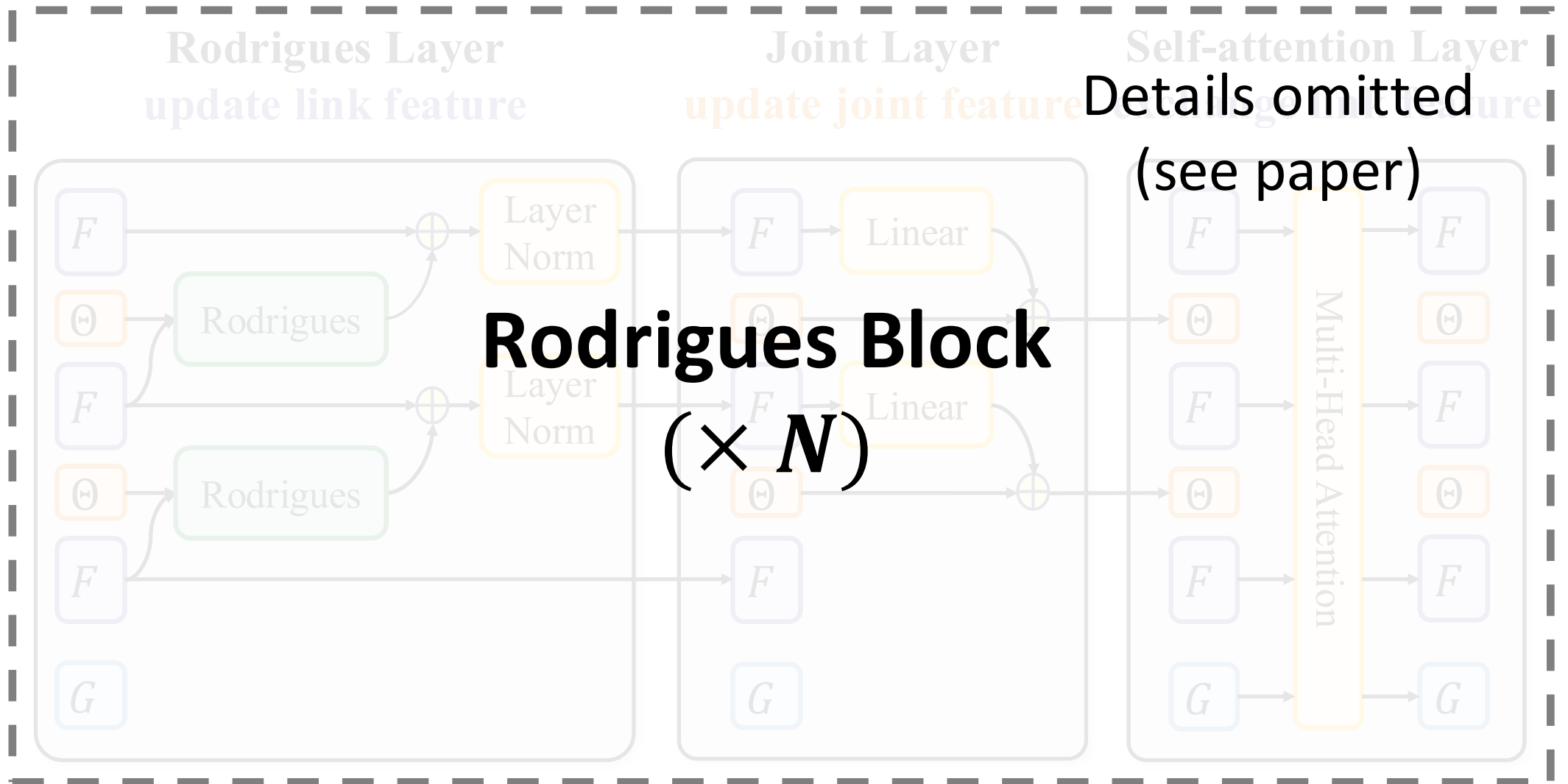
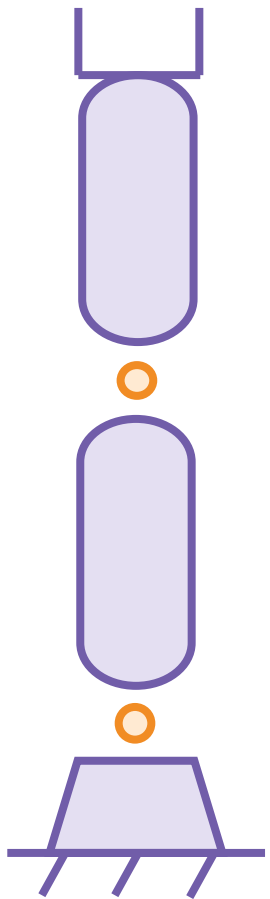
Joint Layer
update joint feature



Self-attention Layer
exchange link feature



Build Rodrigues Network



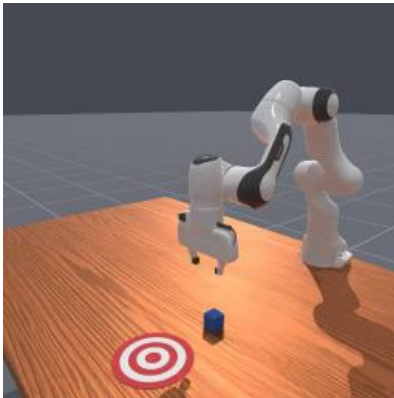
Useful in **applications?**



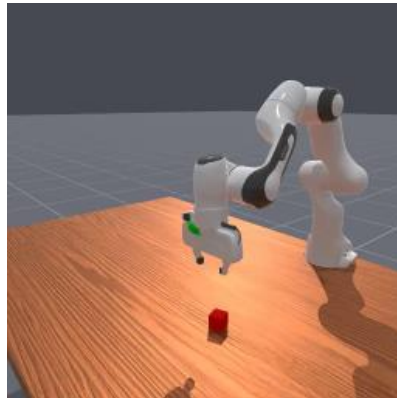
Application 1: Robot Manipulation with IL

- **Benchmark: Derived from ManiSkill**

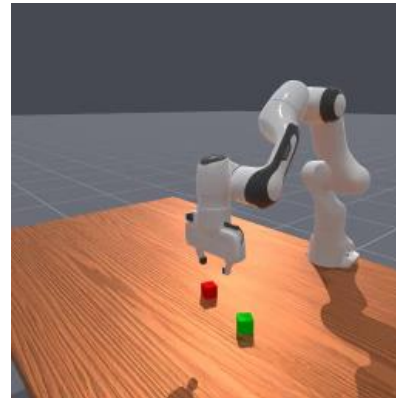
PushCube



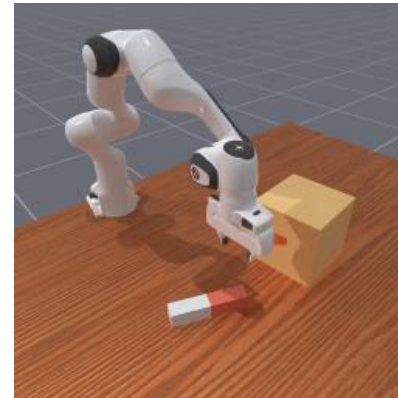
PickCube



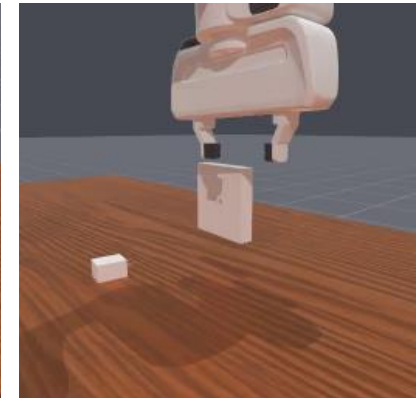
StackCube



PegInserionSide



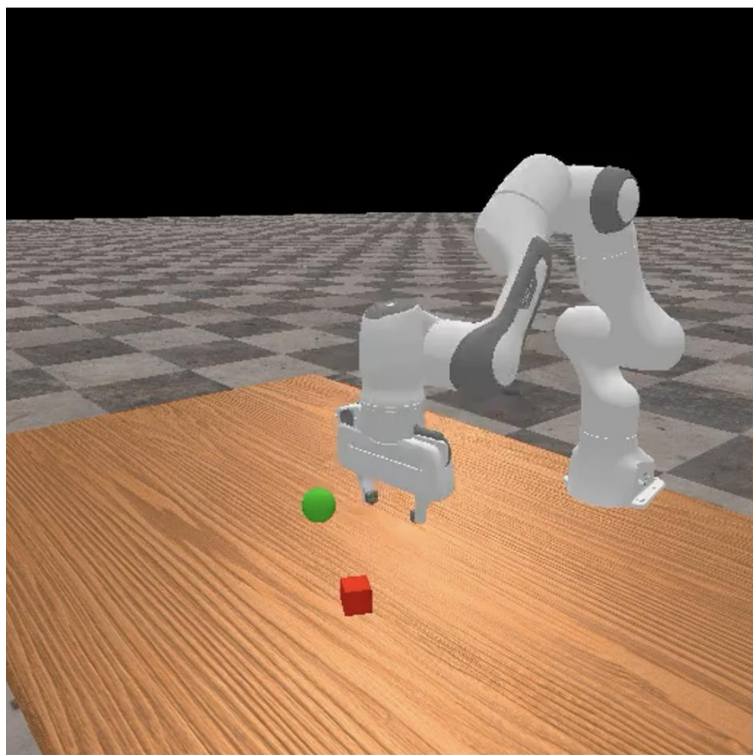
PlugCharger



Images: https://maniskill.readthedocs.io/en/latest/tasks/table_top_gripper/index.html

Kinematics-Inspired Inductive Bias in Manip.

PickCube

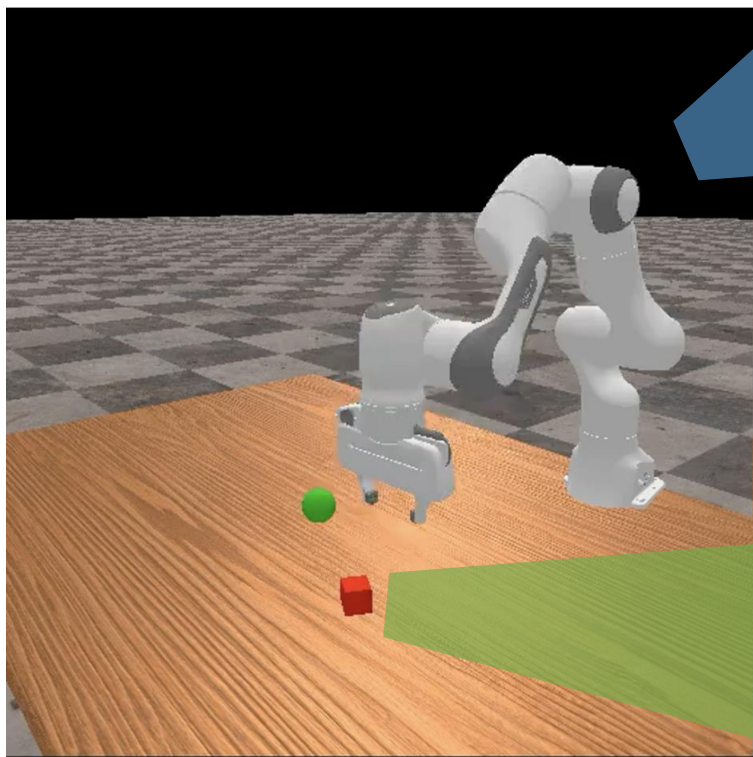


Move **box**
to **goal**



Kinematics-Inspired Inductive Bias in Manip.

PickCube



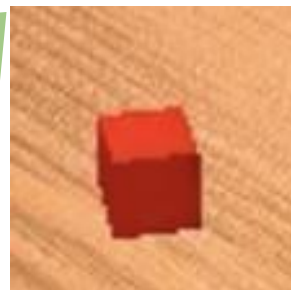
$$\begin{bmatrix} \theta_1 \\ \vdots \\ \theta_7 \end{bmatrix}$$

Action Space

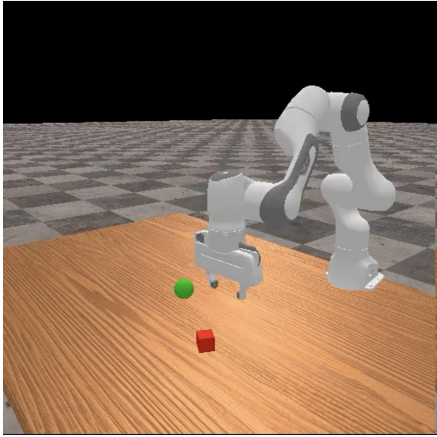
Cross-reason



World Space

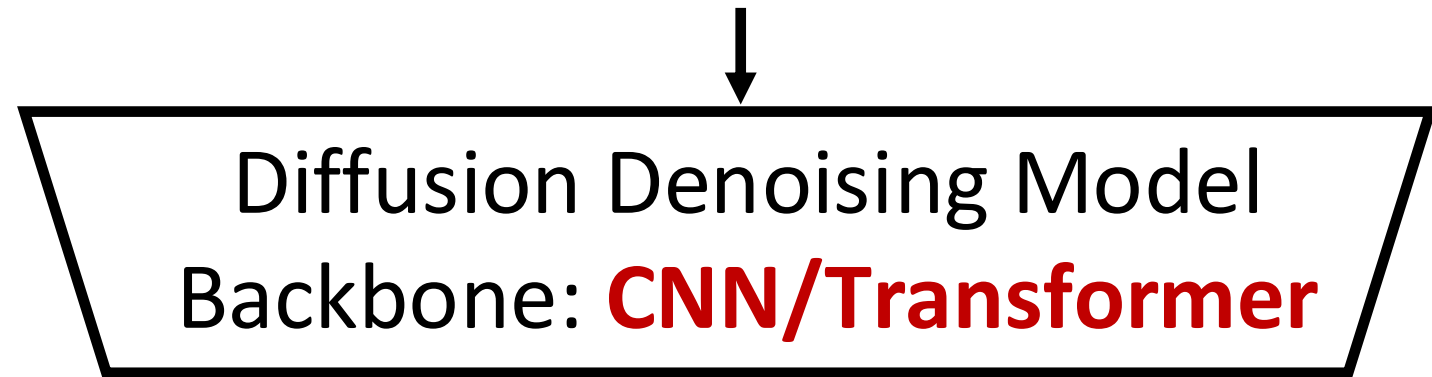
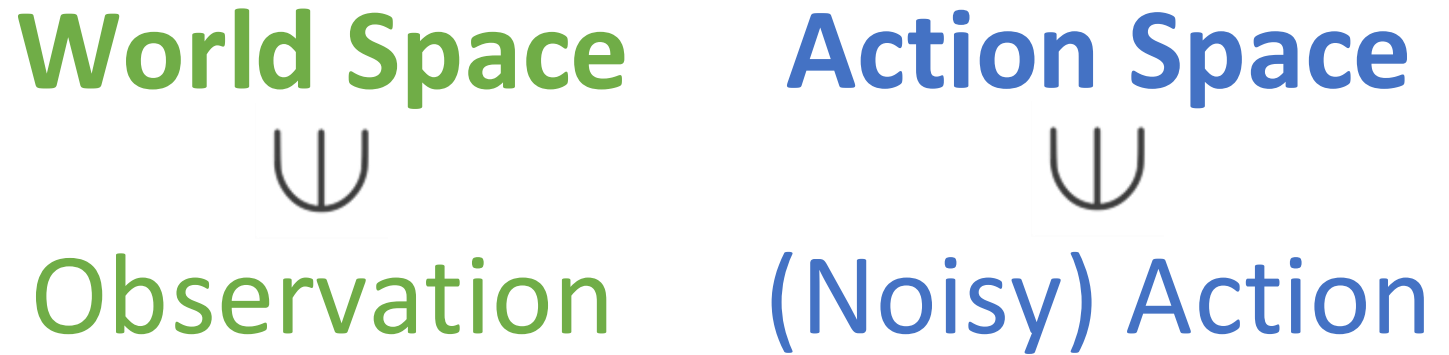


Add Inductive Bias to Established Algo.: DP



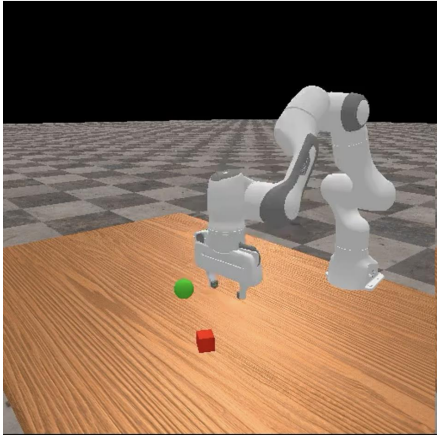
**Without
inductive bias**

[RSS 2023]
Diffusion Policy



Predicted Action Noise

Add Inductive Bias to Established Algo.: DP



World Space



Observation

Action Space



(Noisy) Action



Replace with RodriNet

Diffusion Denoising Model
Backbone: CNN/Transformer



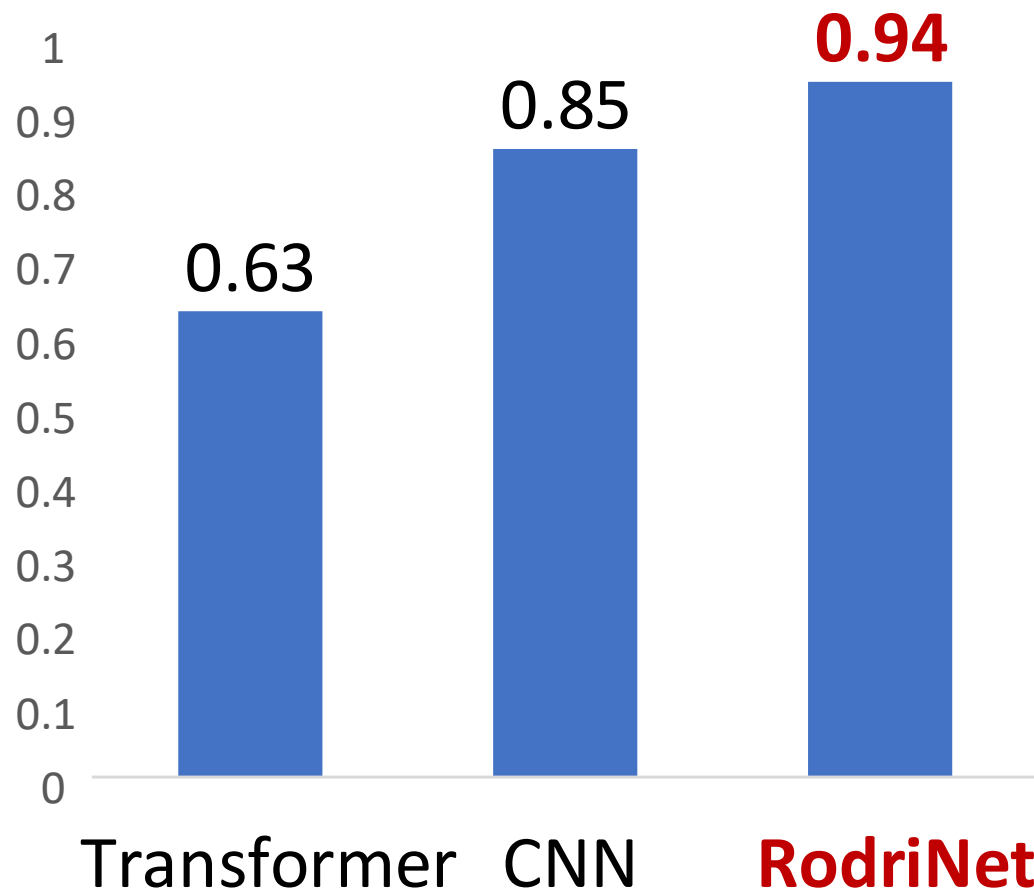
Predicted Action Noise

With
inductive bias

[RSS 2023]
Diffusion Policy

Application 1: Robot Manipulation with IL

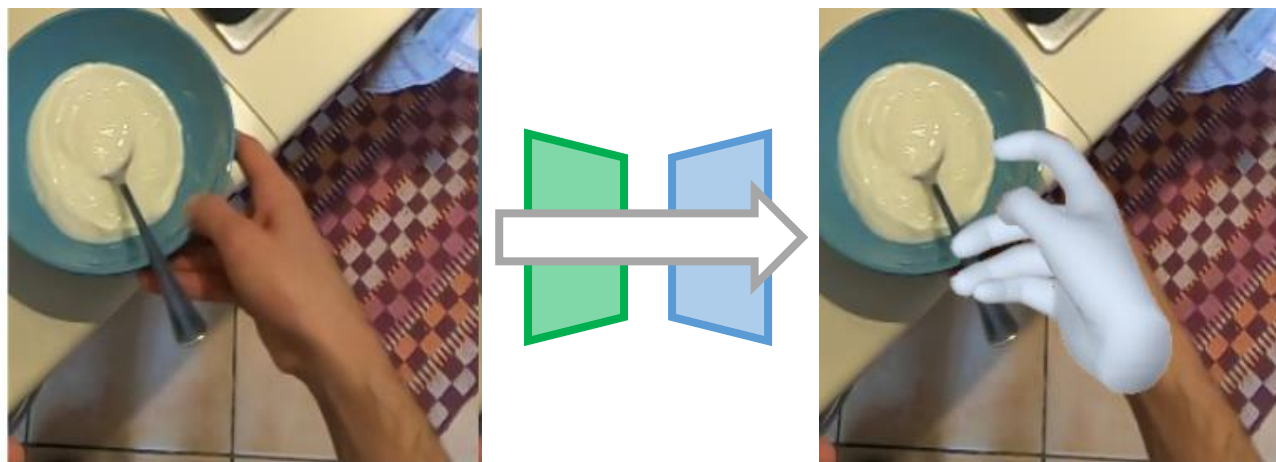
PickCube Sim Succ.



**Learns better from
same demo trajectories**

(see more results in paper)

Application 2: 3D Hand Reconstruction



[CVPR 2024] HaMeR, **transformer**-based

Replace hand pose decoder ↓

Ours, **Rodrigues**-based

(details in paper)

Application 2: 3D Hand Reconstruction

Method	PA-MPJPE ↓	PA-MPVPE ↓	F@5 ↑	F@15 ↑
HaMeR	6.0	5.7	0.785	0.990
HaMeR (Reproduced)	6.2	5.9	0.774	0.989
Ours	5.9	5.6	0.793	0.991

Also works on articulated characters

Summary

- **Neural Rodrigues Operator**
 - Kinematics-inspired inductive bias
- **Rodrigues Network**
 - Embodiment-aware architecture
- **Applicable to diverse action-learning tasks**

Limitations / Future Works

- **Joint space controller**

- Current VLA: gripper + EE space control

- **RL applications**

Rodrigues Network for Learning Robot Actions

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Thanks for Listening!



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GitHub