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Manual-PA: Learning 3D Part Assembly from Instruction Diagrams



Jiahao Zhang¹



Anoop Cherian²



Cristian Rodriguez³



Weijian Deng¹



Stephen Gould¹

¹The Australian National University

²Mitsubishi Electric Research Labs

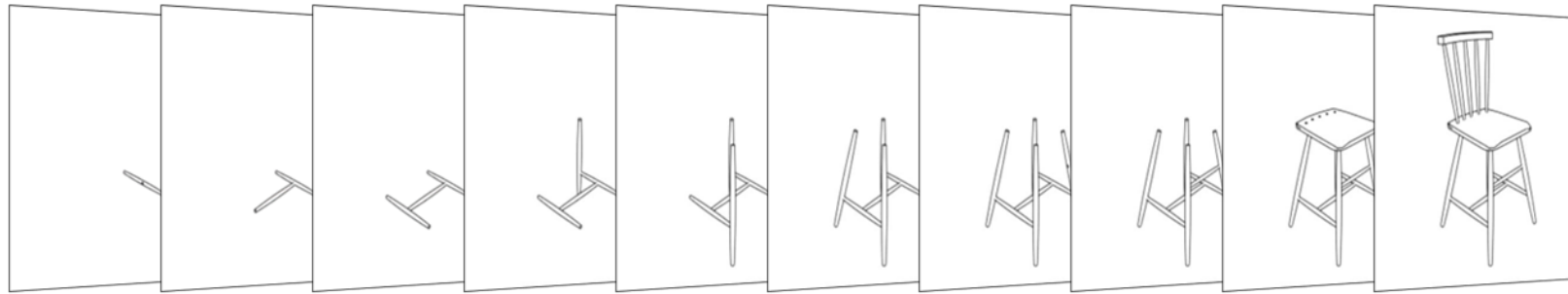
³The Australian Institute for Machine Learning

Code & Dataset: <https://github.com/DavidZhang73/Manual-PA>

Poster: Session 2 (Oct 21 PM)



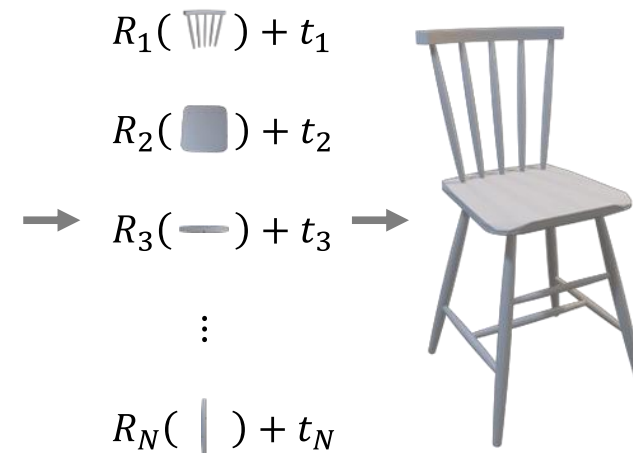
Problem Formulation



(a)

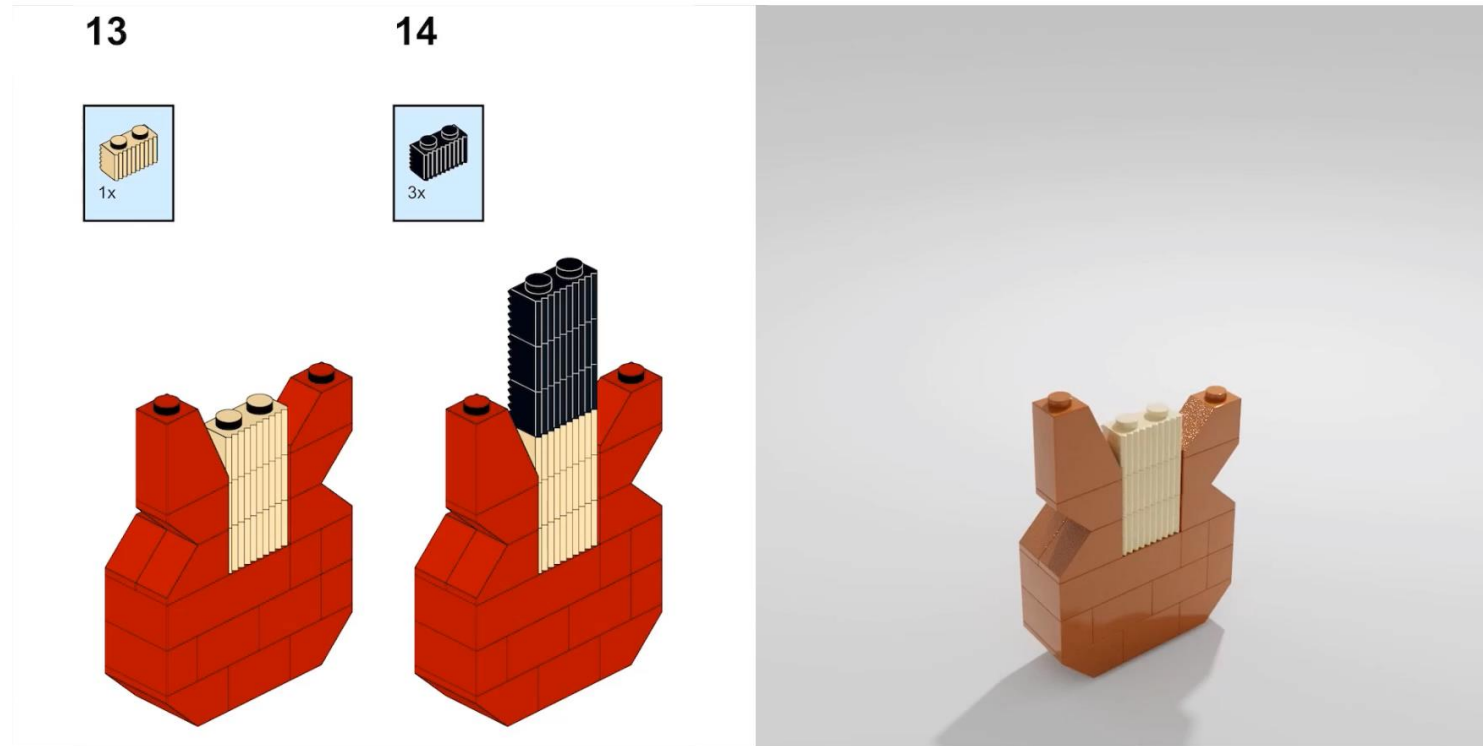


(b)



(c)

Related Works – Assembly Manual – MEPNet

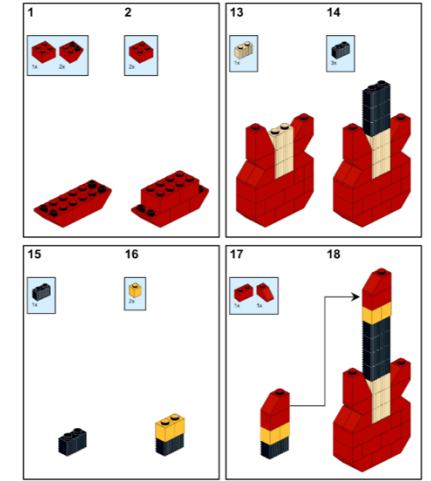
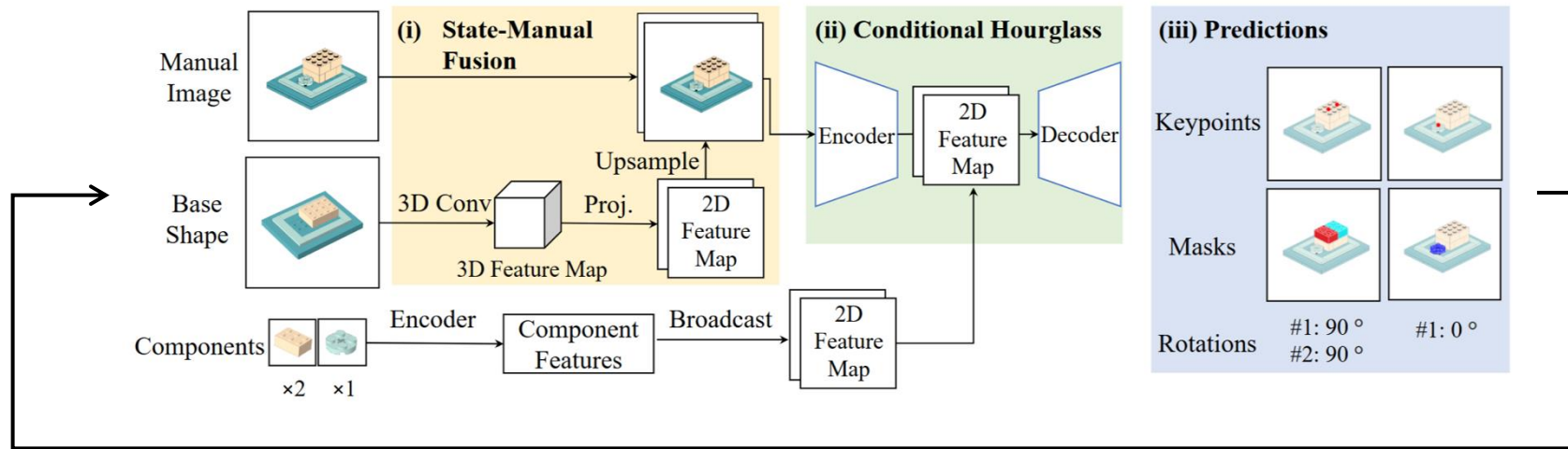


(a) Step-by-step LEGO Manual

(b) Inferred Assembly Process

Our Model can assemble LEGO objects according to manuals.

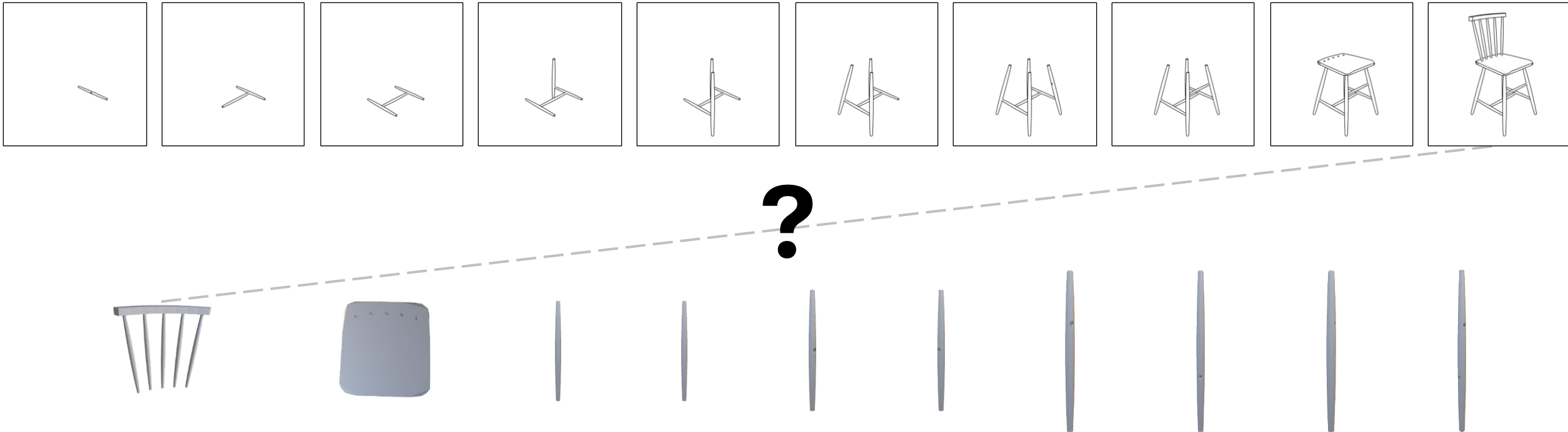
Accumulating Errors!



- Autoregressive arch.
 - Base shape => predicted result from pervious step
- Components => from ground truth

Key Issues

- How to learn correspondence between step diagrams and 3D parts?



Key Issues

- How to learn correspondence between step diagrams and 3D parts?
 - Contrastive learning

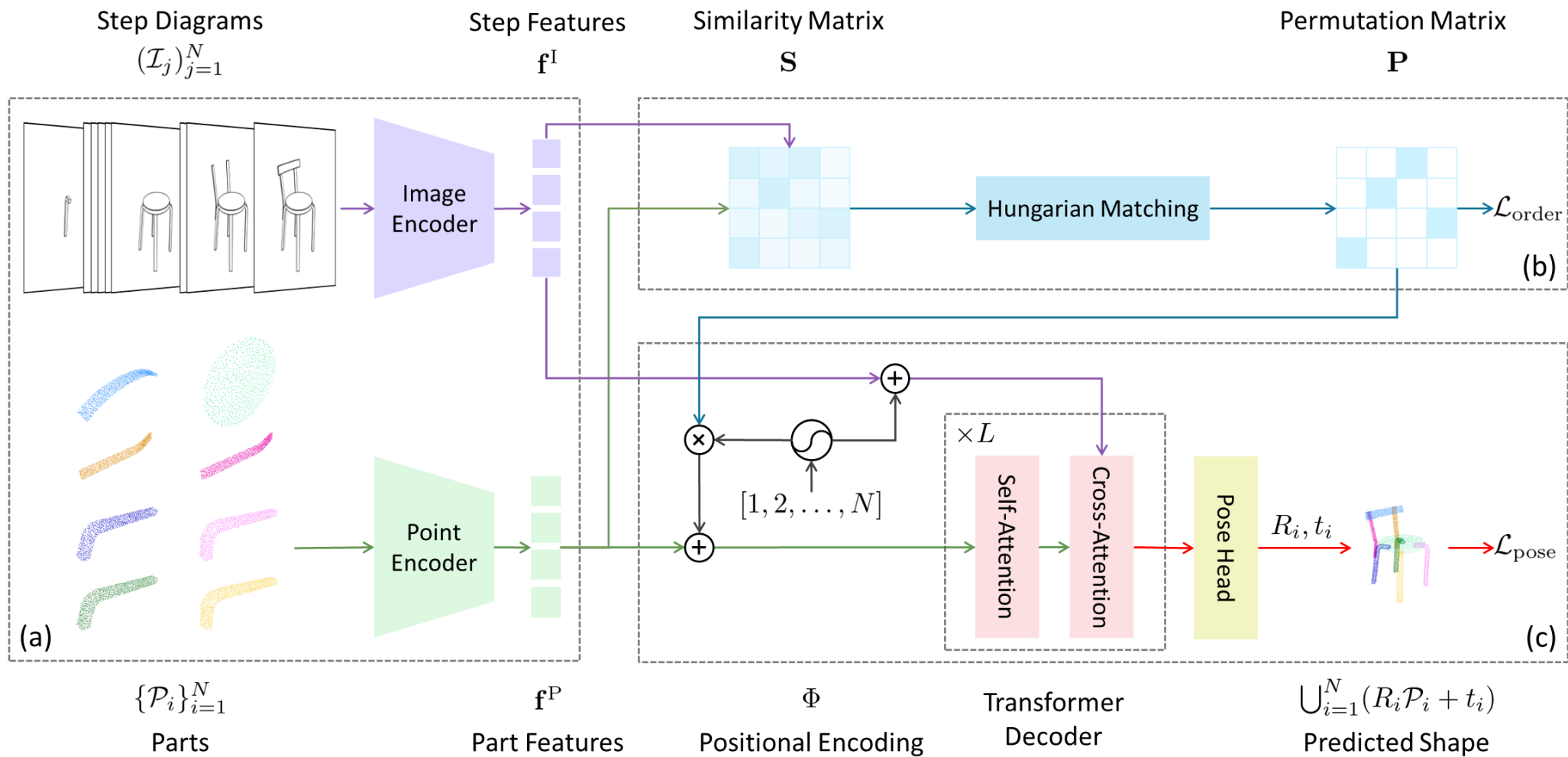
Key Issues

- How to learn correspondence between step diagrams and 3D parts?
 - Contrastive learning
- How to incorporate with the learned correspondence/order for pose estimation?

Key Issues

- How to learn correspondence between step diagrams and 3D parts?
 - Contrastive learning
- How to incorporate with the learned correspondence/order for pose estimation?
 - Via positional encoding as a soft guidance

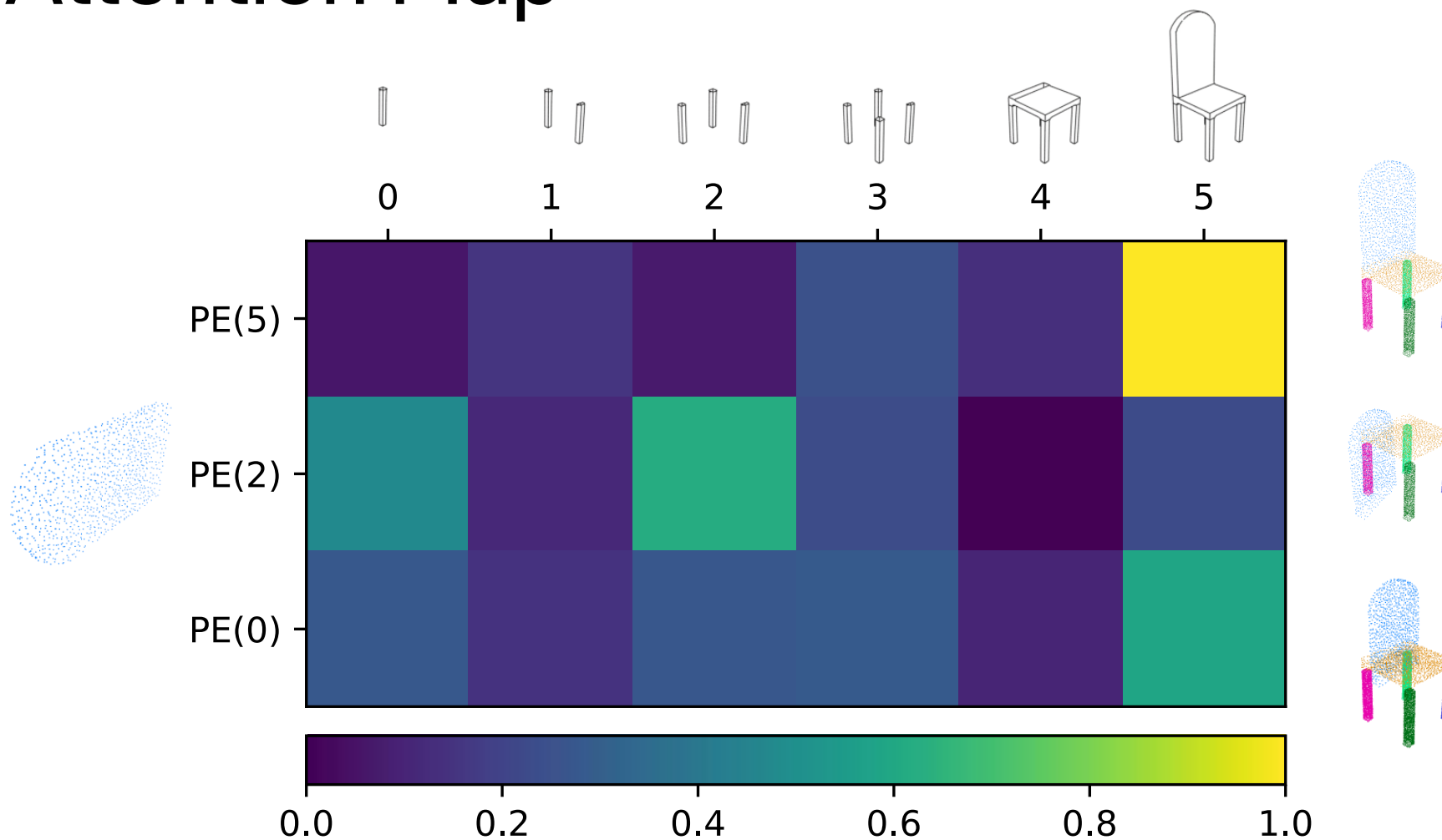
★ Ours: Manual-PA



Soft Guidance?!

- Hard guidance?
 - The 3D assembly model can **NOT** self-correct if the predicted correspondence is wrong.

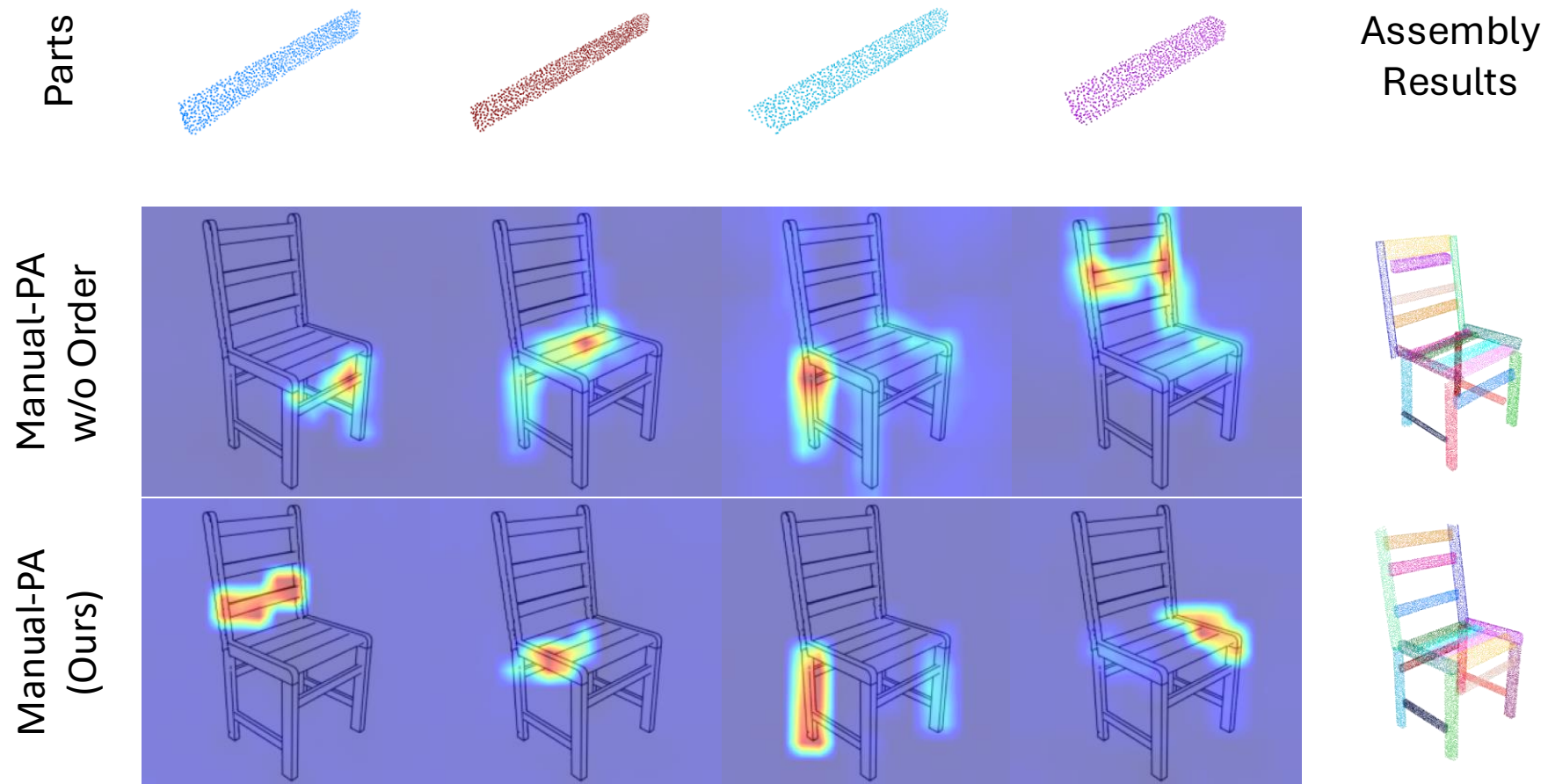
★ Attention Map



Soft Guidance?!

- Hard guidance?
 - The 3D assembly model can **NOT** self-correct if the predicted correspondence is wrong.
- No guidance?
 - The 3D assembly model need to learn the correspondence implicitly by itself.

★ Attention Map

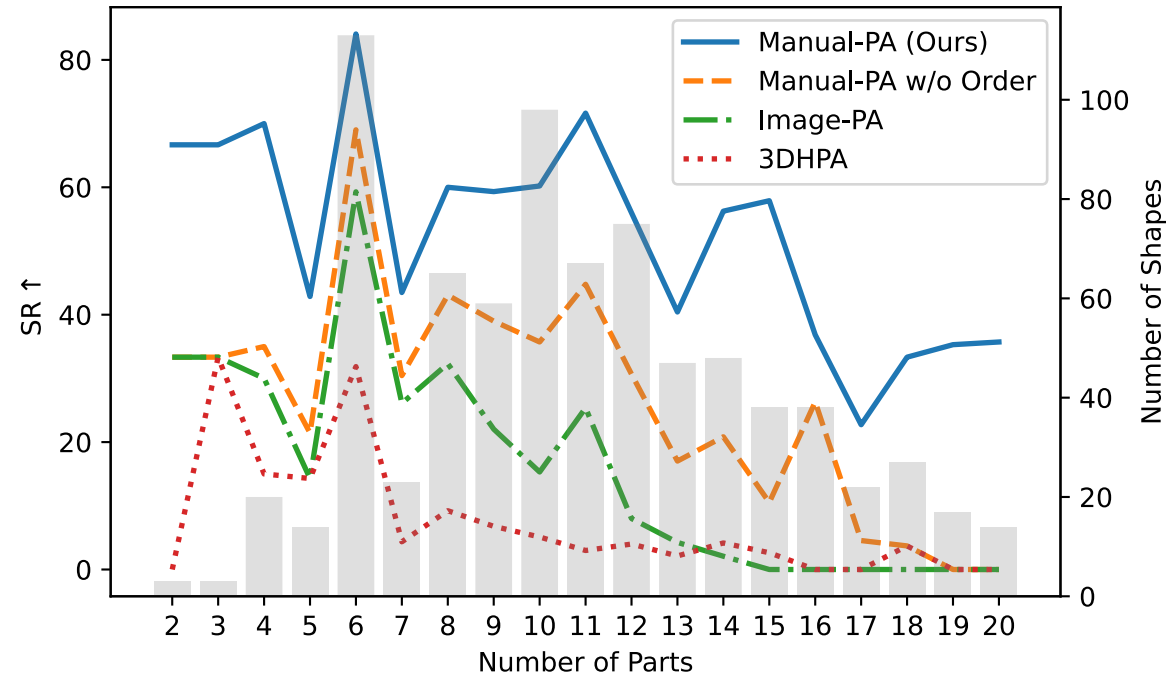


Quantitative Results

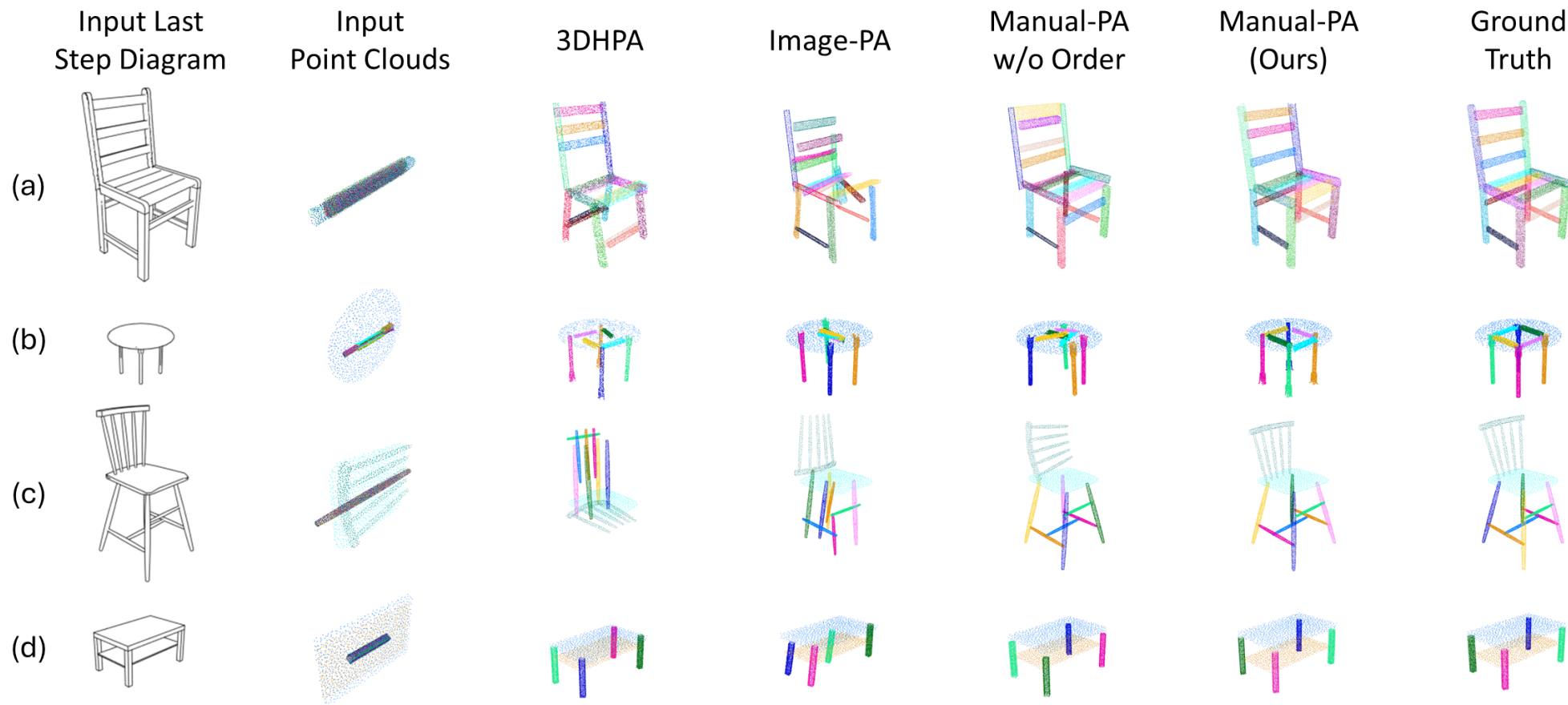
Table 1. 3D part assembly results on the PartNet test split and Ikea-Manual. †: We re-trained the Image-PA model using diagrams (2D line drawing images) as the conditioning input instead of the original RGB images. The values in bold represent the best results, while underlined indicate the second.

Method	Condition	SCD↓		PA↑		SR↑	
		Chair	Table	Chair	Table	Chair	Table
Fully-Supervised on PartNet [26]							
DGL _{NIPS} '20 [52]	-	9.1	5.0	39.00	49.51	-	-
IET _{RA-L} '22 [55]	-	5.4	3.5	62.80	61.67	-	-
Score-PA _{BMVC} '23 [8]	-	7.4	4.5	42.11	51.55	8.320	11.23
CCS _{AAAI} '24 [56]	-	7.0	-	53.59	-	-	-
3DHPA _{CVPR} '24 [9]	-	5.1	<u>2.8</u>	64.13	64.83	-	-
RGL _{WACV} '22 [27]	Sequence	8.7	4.8	49.06	54.16	-	-
SPAFormer _{ArXiv} '24 [51]	Sequence	6.7	3.8	55.88	64.38	16.40	33.50
Joint-PA _{CVPR} '24 [24]	Joint	6.0	7.0	72.80	67.40	-	-
Image-PA _{ECCV} '20 [22]	Image	6.7	3.7	45.40	71.60	-	-
Image-PA _{ECCV} '20 [†]	Diagram	5.9	3.9	62.67	70.10	19.97	32.83
Manual-PA w/o Order	Manual	<u>3.0</u>	3.6	<u>79.07</u>	<u>74.03</u>	<u>34.13</u>	<u>37.71</u>
Manual-PA (Ours)	Manual	1.7	1.8	89.06	87.41	58.03	56.66
Zero-Shot on IKEA-Manual [47]							
3DHPA _{CVPR} '24	-	34.3	37.8	1.914	4.027	0.000	0.000
Image-PA _{ECCV} '20 [†]	Diagram	17.3	14.7	19.07	36.74	0.000	<u>10.53</u>
Manual-PA w/o Order	Manual	<u>12.8</u>	<u>8.9</u>	<u>38.36</u>	<u>42.01</u>	<u>1.754</u>	<u>10.53</u>
Manual-PA (Ours)	Manual	11.4	4.8	42.51	49.72	3.509	15.79

Quantitative Results



Qualitative Results

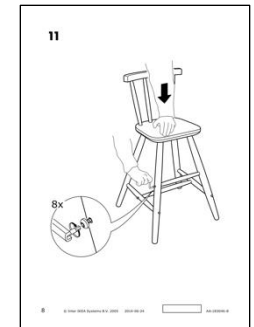
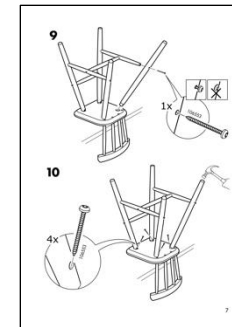
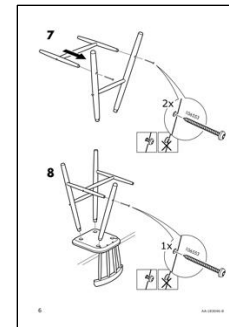
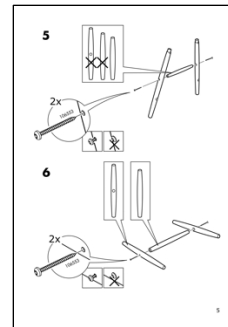
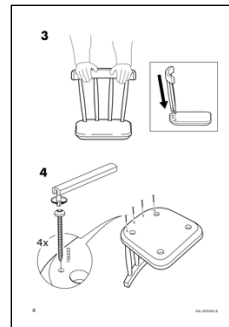
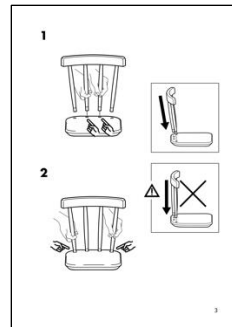
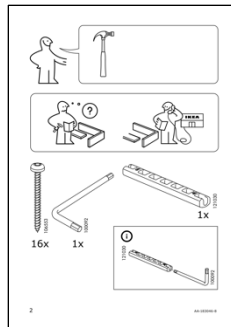
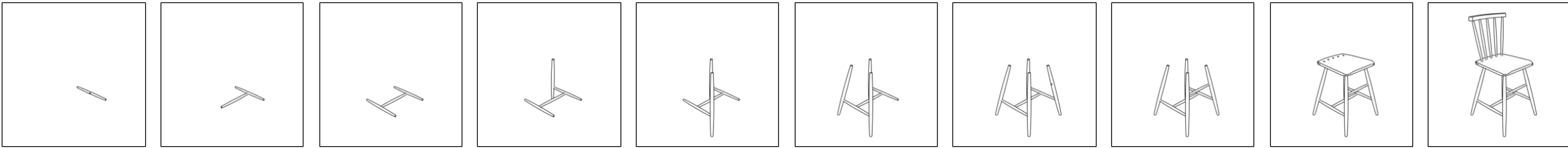


Limitations and Future Works

- Category-level vs. Universal

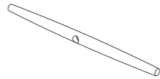
Limitations and Future Works

- Synthetic vs. Real World Manual



Demonstration Video

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Manual

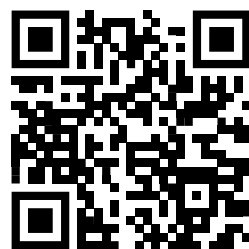


Image-PA



Manual-PA (Ours)

Thanks!



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