



3D Interacting Hand Pose Estimation by Hand De-occlusion and Removal

Hao Meng^{1,3*} Sheng Jin^{2,3*} Wentao Liu^{3,4} Chen Qian³

Mengxiang Lin¹ Wanli Ouyang^{4,5} Ping Luo²

¹ Beihang University

² The University of Hong Kong

³ SenseTime Research and Tetras.AI

⁴ Shanghai AI Lab

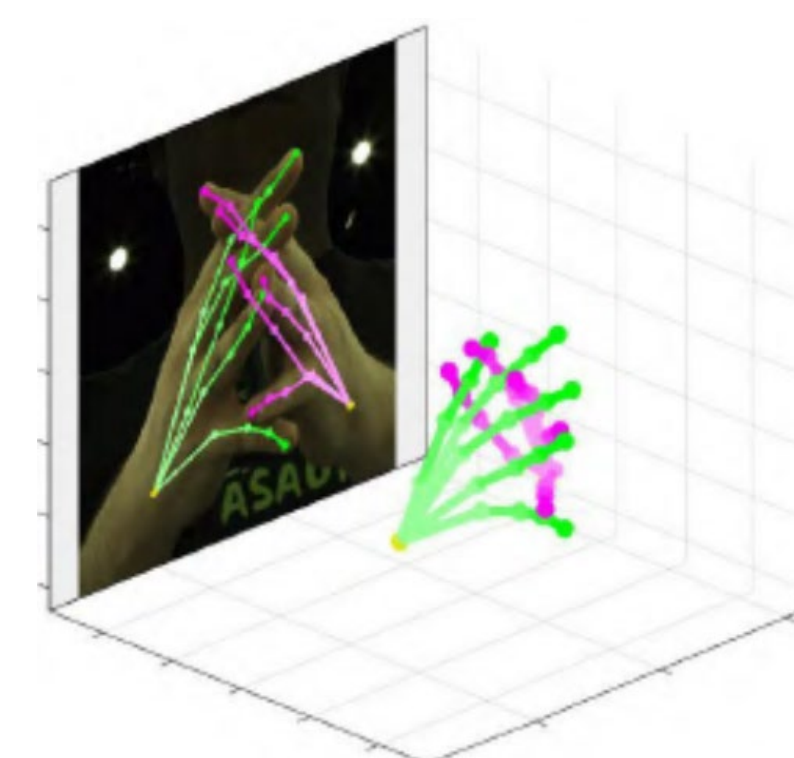
⁵ The University of Sydney



Task

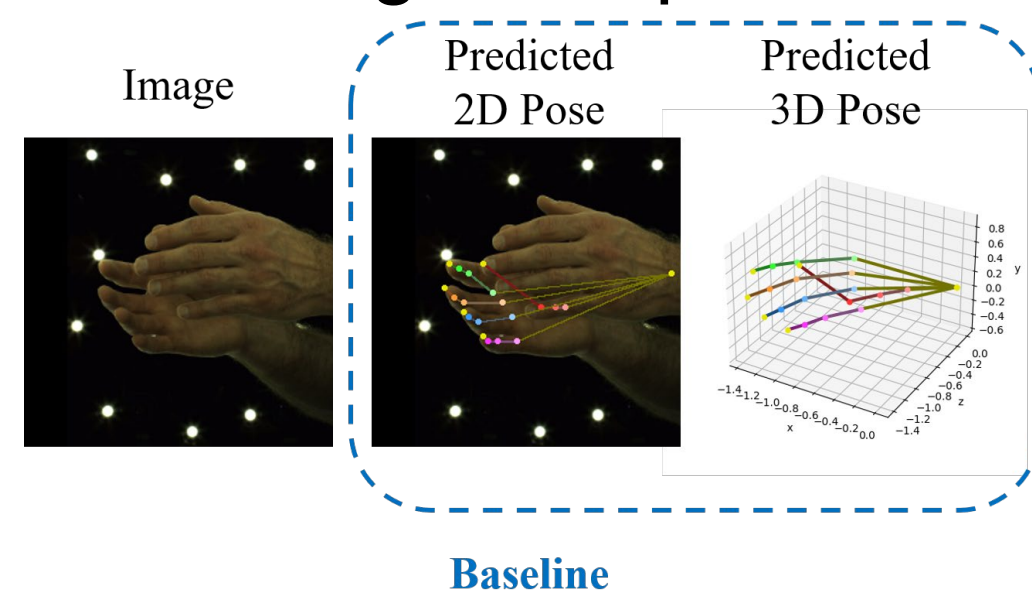
3D Interacting Hand Pose Estimation

Estimating 3D keypoint localization of two interacting hands.



Motivation

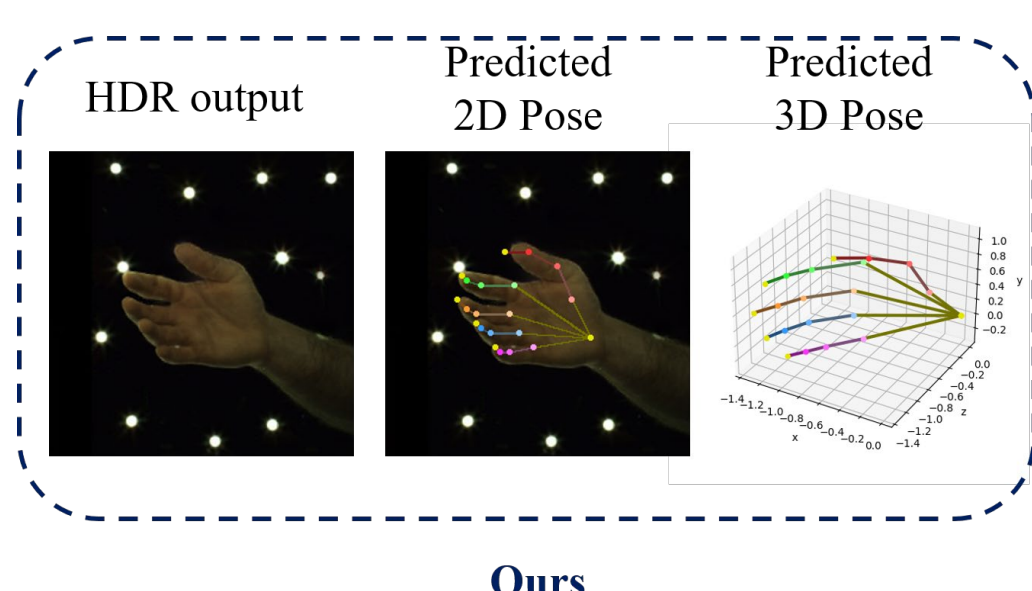
Interacting hand pose estimation VS single hand pose estimation:



☹️ Occlusion between two hands

☹️ Ambiguity between two hands

HDR (Hand De-occlusion and Removal):



😊 Single hand pose estimation for every hand

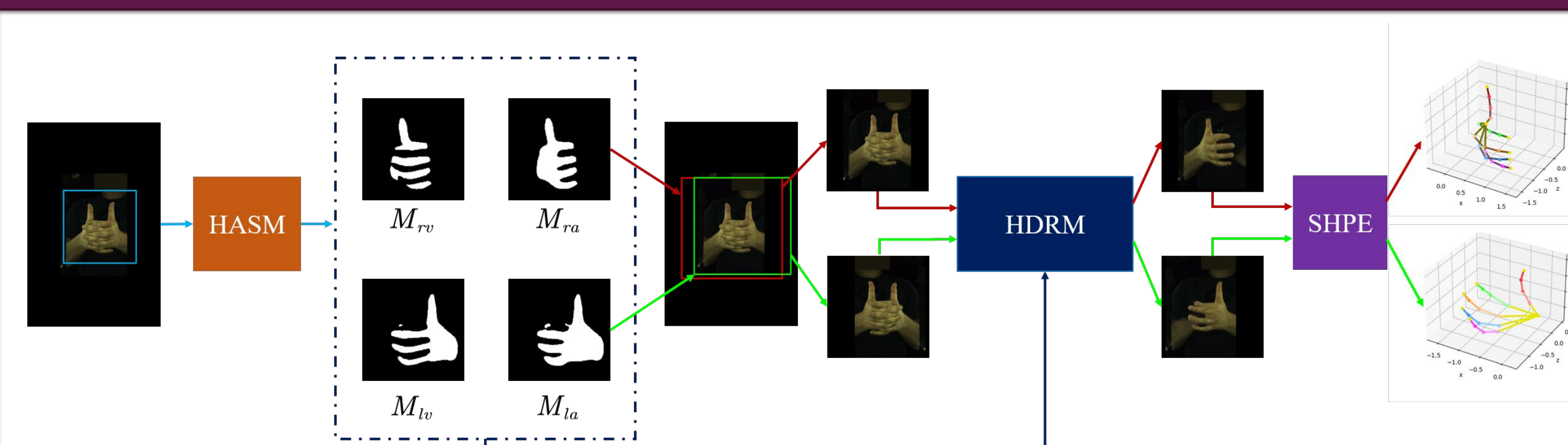
😊 De-occlusion: restore pixel values of occluded part of hand

😊 Removal: remove the other hand, i.e., predict pixel values of background behind the other hand

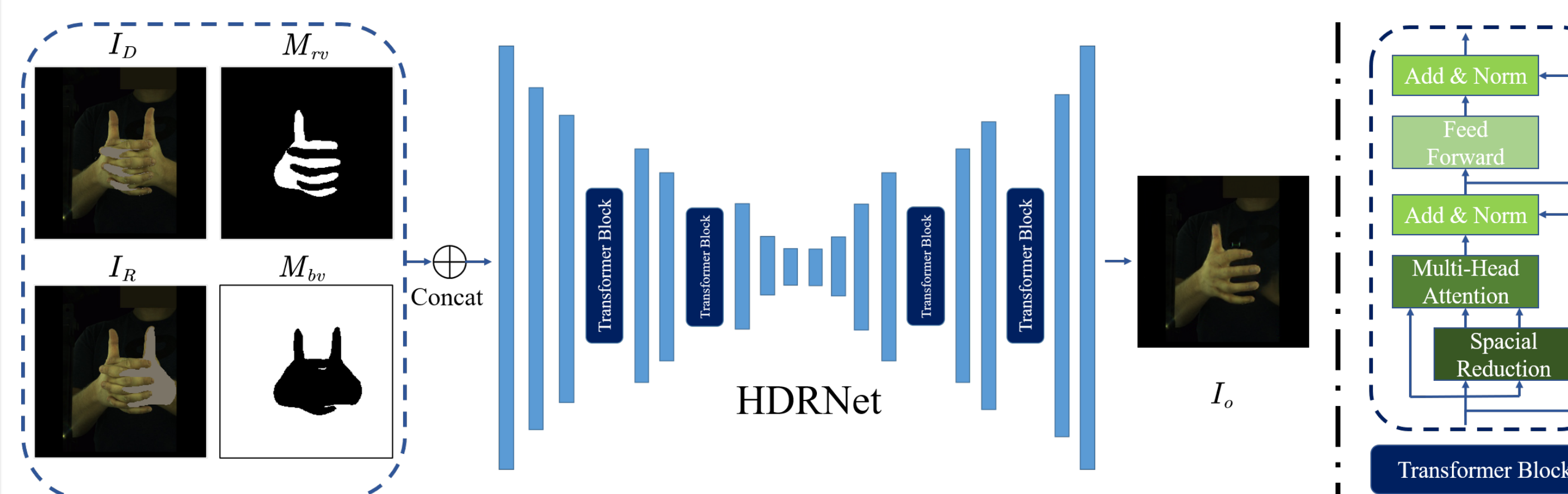
Contributions

- We propose a novel Hand De-occlusion and Removal (HDR) framework to tackle the challenging task of 3D interacting hand pose estimation.
- We propose to explicitly handle the challenges of self-occlusion by hand de-occlusion and the homogeneous appearance ambiguity by distractor removal. To the best of our knowledge, we are the first to apply de-occlusion techniques to improve the downstream pose estimation accuracy.
- We propose the first large-scale synthetic Amodal InterHand Dataset (AIH) to settle the task of hand de-occlusion and removal. We envision that AIH will foster the development of the related research.

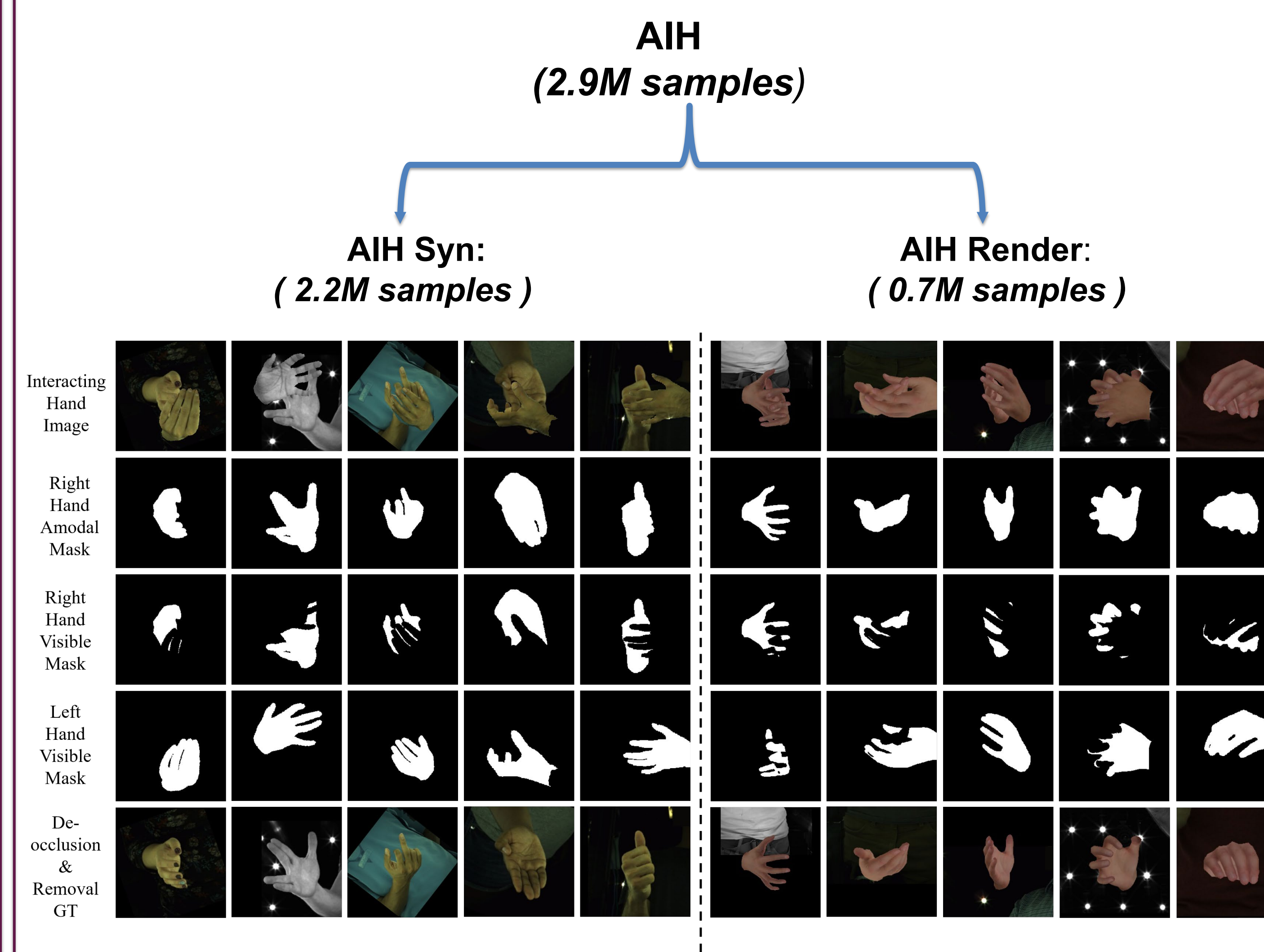
Overview



Hand De-occlusion and Removal Module (HDRM)



Amodal InterHand (AIH) Dataset



Experiment Results

Quantitative results

Comparisons on InterHand2.6M V1.0 Dataset

Methods	InterHand2.6M - ALL branch				InterHand2.6M - M branch		
	IH26M-SH	IH26M-IH	IH26M-ALL	IH26M-Inter	IH26M-SH	IH26M-IH	IH26M-ALL
*Boukhayma <i>et al.</i> [4]	-	-	27.14	31.46	-	-	-
*Pose2Mesh [5]	-	-	27.10	32.11	-	-	-
*BiHand [35]	-	-	25.10	28.23	-	-	-
*Rong <i>et al.</i> [27]	-	-	17.12	20.66	-	-	-
DIGIT [7]	-	14.27	-	-	-	-	-
InterNet [21]	12.16	16.02	14.21	18.04	12.52	18.04	15.28
HDR (Ours)	8.51	13.12	10.97	14.74	8.52	14.98	11.74

Comparisons on Tzionas Dataset

Model	Boukhayma <i>et al.</i> [4]	Wang <i>et al.</i> [31]	InterNet [21]	Kim <i>et al.</i> [13]	SHPE [39]	SHPE+HDR
EPE↓	12.91	13.31	17.61	12.42	14.88	8.70

Comparisons on InterHand2.6M V1.0 dataset (ALL branch)

Methods	Train (ALL, IH26M-SH)		Train (ALL, IH26M-SH + AIH)	
	IH26M-IH	IH26M-ALL	IH26M-IH	IH26M-ALL
SHPE [39]	39.96	25.90	30.23	20.93
+HDR (Ours)	25.93	18.39	23.99	17.58

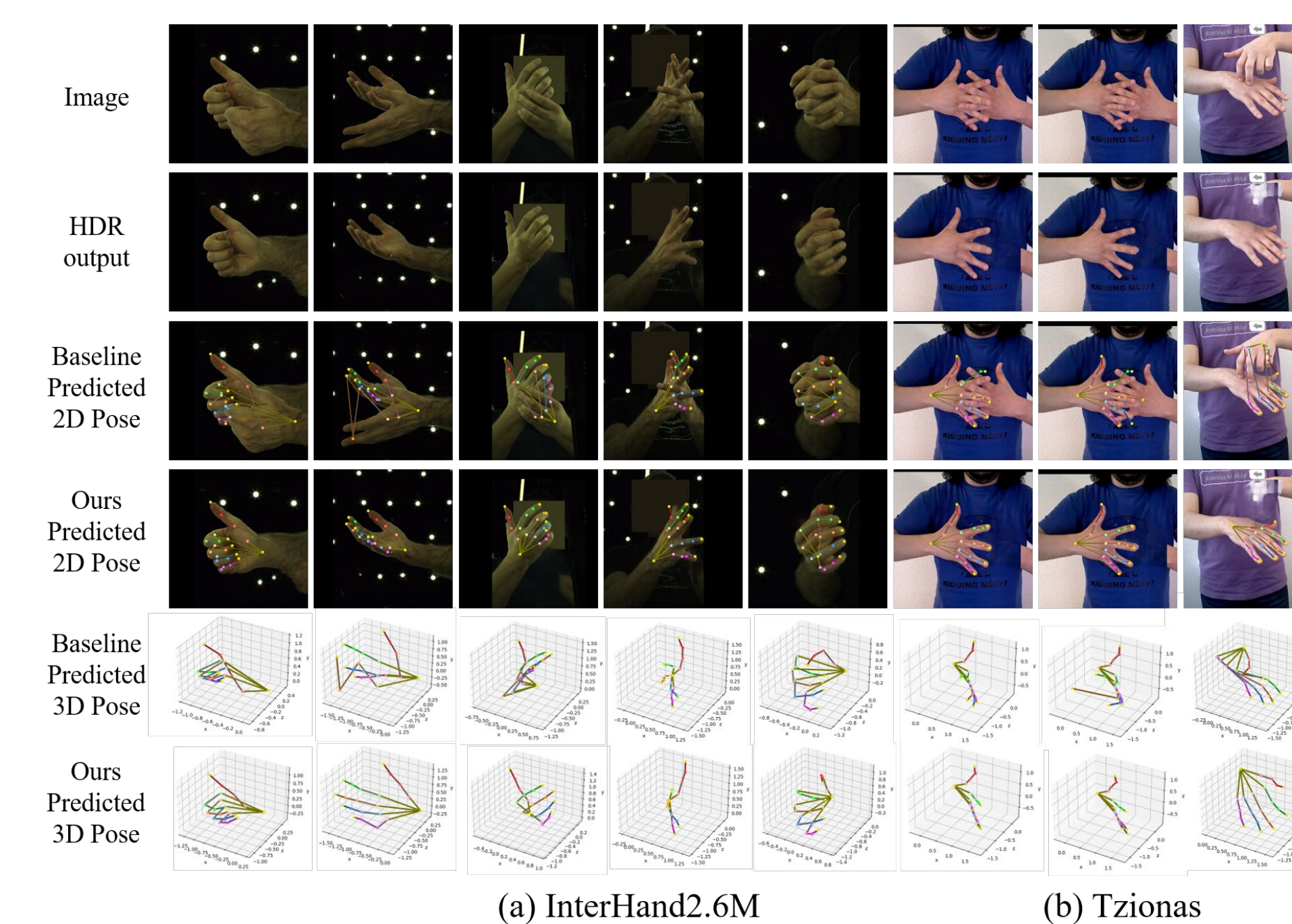
Ablation Studies

#	Methods	MPJPE (mm)	Δ
#1	SHPE [39] only	25.78	+7.80 (43.4%)
#2	w/o Removal	24.16	+6.18 (34.4%)
#3	w/o De-occlusion	19.69	+1.71 (9.5%)
#4	w/o Discriminator	18.11	+0.13 (0.7%)
#5	w/o Transformer Block	18.85	+0.87 (4.8%)
#6	AIH_Render only	18.10	+0.12 (0.7%)
#7	AIH_Syn only	18.35	+0.37 (2.1%)
#8	Ours	17.98	-

Comparisons on InterHand2.6M V1.0 dataset (Machine annot branch)

Methods	Train (M, IH26M-SH)		Train (M, IH26M-SH + AIH)	
	IH26M-IH	IH26M-ALL	IH26M-IH	IH26M-ALL
SHPE [39]	40.98	25.78	32.27	21.66
+HDR (Ours)	25.45	17.98	24.59	17.80

Qualitative results



Time complexity analysis

Test on one Tesla P40 GPU in a single thread

