



ID: THU-PM-053 https://osx-ubody.github.io

One-Stage 3D Whole-body Mesh Recovery with Component Aware Transformer

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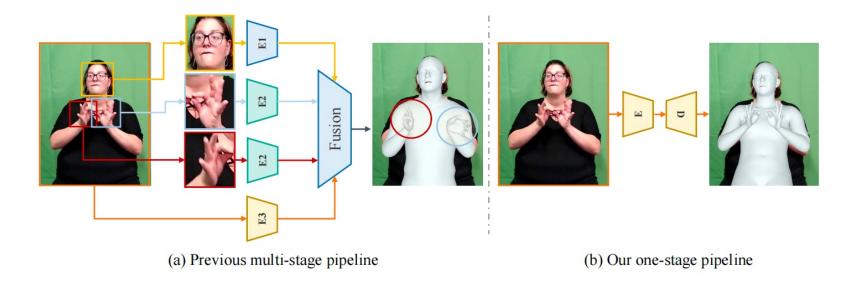
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➤ A large-scale **upper-body dataset**, UBody, bridges the gap between the basic task and downstream applications.

> New state-of-the-art performance on three datasets.

		AGORA-test						EHF					
Method	MPVPE ↓			N-MP	N-MPVPE↓		MPVPE ↓			A-MPVPI	E↓	MPJPE ↓	PA-MPJPE ↓
	All	Hands	Face	All	Body	All	Hands	Face	All	Hands	Face	Body	Body
ExPose [36]	217.3	73.1	51.1	265.0	184.8	77.1	51.6	35.0	54.5	12.8	5.8	93.4	60.7
FrankMocap [41]	-	55.2	-	-	207.8	107.6	42.8	-	57.5	12.6	-	96.7	61.9
PIXIE [13]	191.8	49.3	50.2	233.9	173.4	89.2	42.8	32.7	55.0	11.1	4.6	91.0	61.3
Hand4Whole [29]	-	-	-	-	-	79.2	43.2	25.0	53.1	12.1	5.8	-	-
Hand4Whole [29]×	135.5	47.2	41.6	144.1	96.0	76.8	39.8	26.1	50.3	10.8	5.8	86.6	54.4
OSX (Ours)	122.8 ↓ _{9.5%}	45.7	36.2	130.6	85.3	70.8 ↓ _{7.8%}	53.7	26.4	<u>48.7</u>	15.9	6.0	74.7 ↓ _{13.4%}	45.1

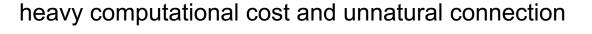
Background

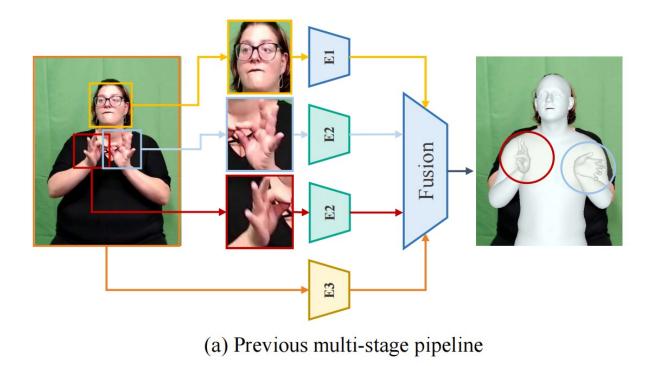


Input

Result

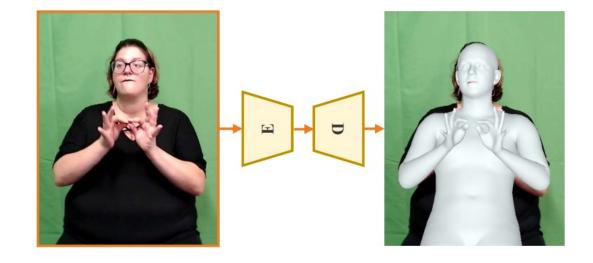
➤ Multi-stage methods: use three separate network for the body pose, facial expression and hand pose estimation.





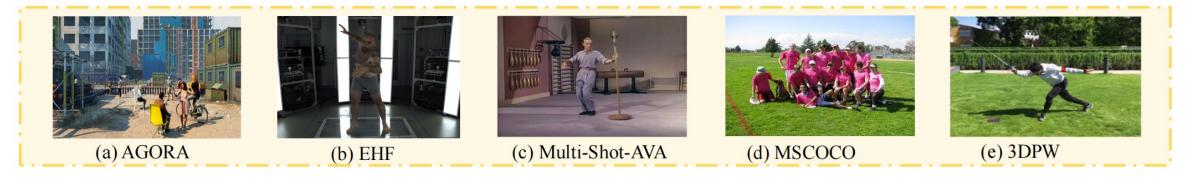
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➤ Data discrepancy: Existing datasets are whole-body scenes. But in many daily life scenes, upper body is a major focus.

Model can not perform well in real-life scenes



Existing Datasets

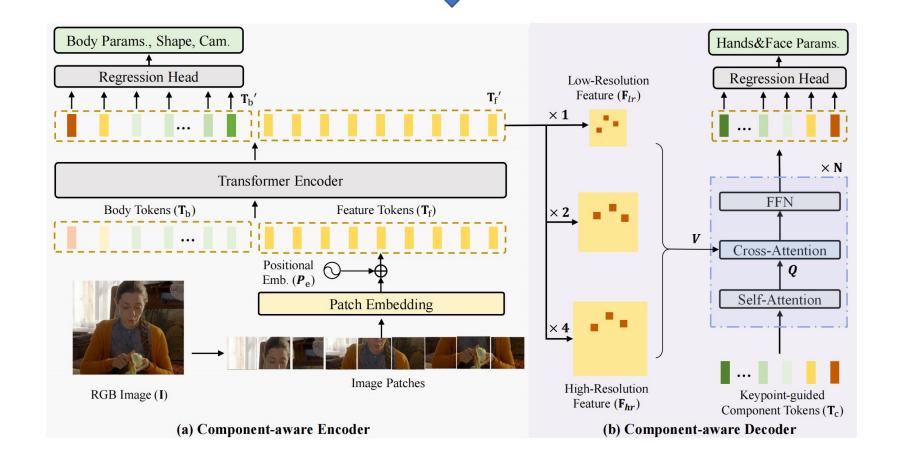
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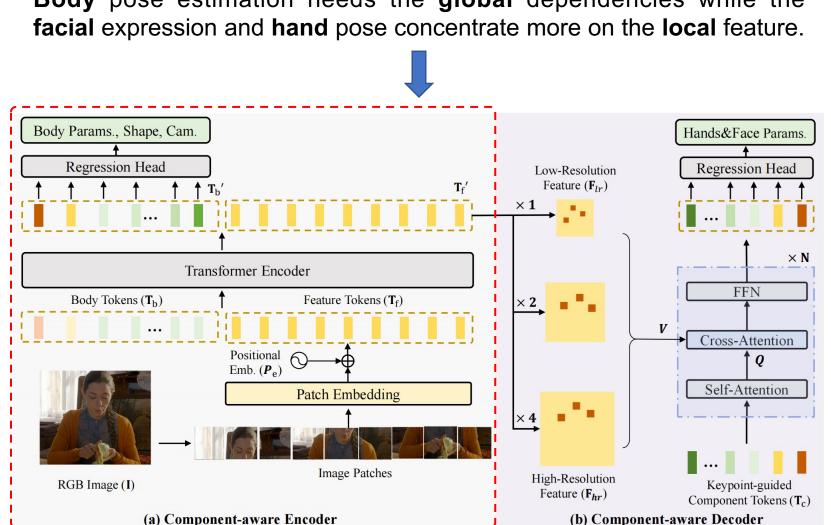
Upper-Body Dataset?



Real-life Scenes

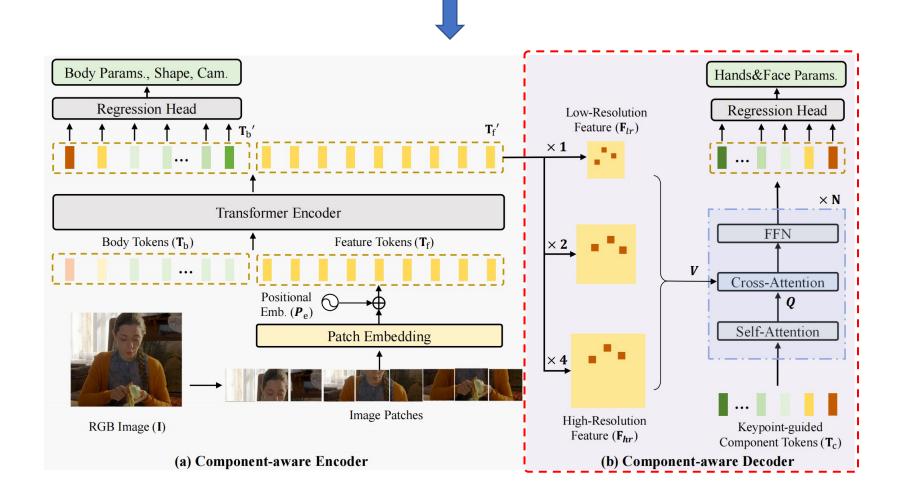
Body pose estimation needs the **global** dependencies while the **facial** expression and **hand** pose concentrate more on the **local** feature.



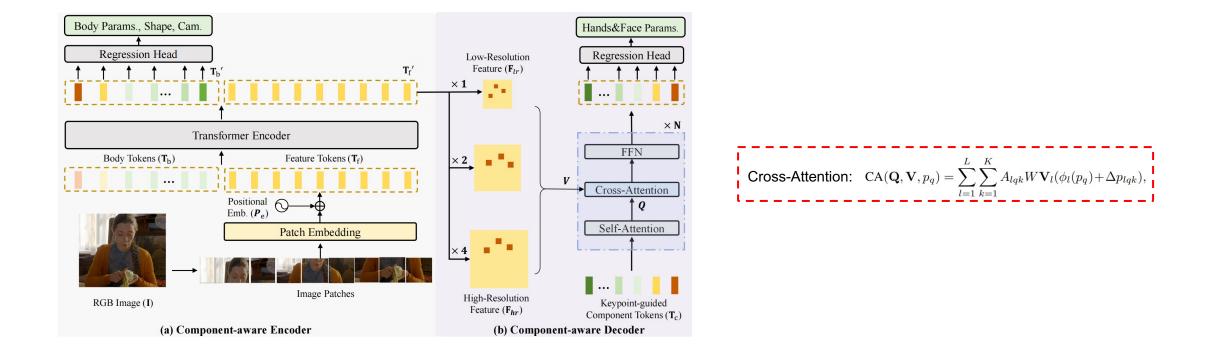


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Dataset Characteristics

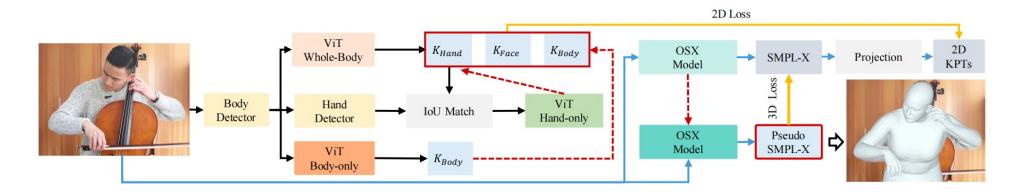
- 1. Upper-body: with partial observation and heavy truncation
- 2. Expressive: rich hand gestures and facial expressions
- 3. Large-scale: fifteen real-life scenarios, more than 1M frames
- 4. Challenging: severe truncation, dynamic camera view, etc.

➤ 15 Real-Life Scenes



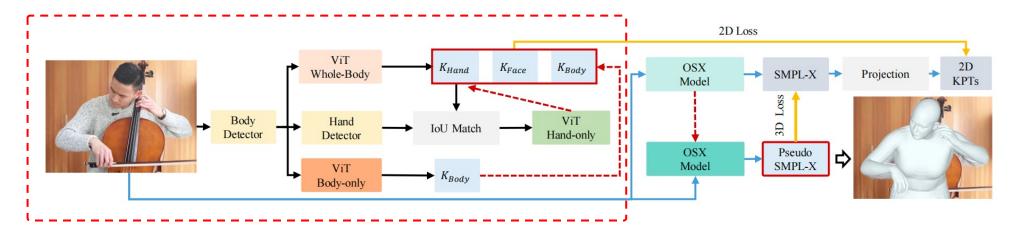
UBody contains 15 human-centric real-life scenes, which mainly focus on the upper-body parts.

► Annotation Pipeline



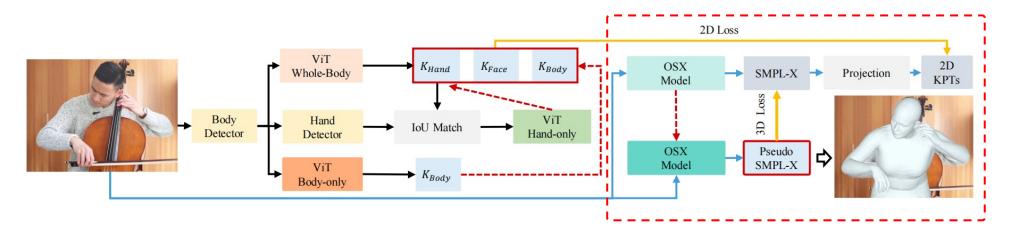
We annotate 2d whole-body keypoints and 3d whole-body mesh with an automatic annotation pipeline.

► Annotation Pipeline



2D whole-body **keypoints** annotation: BodyHands detects bounding box, 4 ViT-based models predict the whole-body, body, face and hand keypoints.

► Annotation Pipeline



3D SMPLX fitting: Iterative training-labeling-revision loop to fit the projected 2D keypoints into the annotated 2D keypoints.

➤ Annotation Visualization



		EHF						3DPW					
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FrankMocap [56]	-	55.2	<u>_</u>	-	207.8	107.6	42.8	2	57.5	12.6	-	96.7	61.9
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\succ Comparison with existing methods on three datasets.

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➤ Benchmark on the proposed UBody datasets.

	N	IPVPE 、	ł	P A	-MPVP	PA-MPJPE \downarrow		
Method	All	Hand	Face	All	Hand	Face	Body	Hand
ExPose [48]	171.5	83.7	45.1	66.9	12.0	3.9	70.7	12.3
PIXIE [16]	168.4	55.6	45.2	61.7	12.2	4.2	66.8	12.3
Hand4Whole [39]	104.1	<u>45.7</u>	27.0	44.8	<u>8.9</u>	2.8	45.5	<u>9.0</u>
Hand4Whole [39] \times	157.4	62.2	49.8	82.2	9.8	3.9	92.8	10.0
OSX (Ours) OSX (Ours)†	92.4 81.9	47.7 41.5	<u>24.9</u> 21.2	$\frac{42.4}{42.2}$	10.8 8.6	$\frac{2.4}{2.0}$	<u>42.9</u> 48.4	11.0 8.8

		AGORA-test						EHF					
Method	MPVPE ↓			N-MPVPE↓		MPVPE ↓			PA-MPVPE↓			MPJPE↓	PA-MPJPE ↓
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FrankMocap [56]	2	55.2	1	-	207.8	107.6	42.8	2	57.5	12.6	2	96.7	61.9
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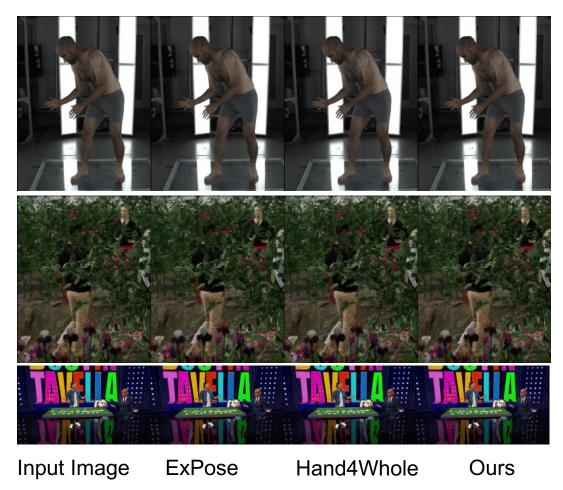
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OSX (Ours)	92.4	47.7	<u>24.9</u>	42.4	10.8	<u>2.4</u>	42.9	11.0
OSX (Ours)†	81.9	41.5	21.2	42.2	8.6	2.0	38.4	8.8

➤ Ablation studies of the keypointguided attention and upsample scale.

Hand	Ours	w/o <i>H.D</i> .	w/o <i>K.G</i>	w/o both
MPVPE	53.7	55.3	55.1	56.4
PA-MPVPE	15.9	17.7	17.6	18.1
Face	Ours	w/o <i>F.D</i> .	w/o <i>K.G</i>	w/o both
MPVPE	26.4	27.2	26.4	26.8
PA-MPVPE	6.0	5.9	5.8	6.0
Upsampling	$ \times 1$	× 2	× 4	× 8
MPVPE	54.9	54.3	53.7	54.1

 \succ Visual Comparison with existing methods on three datasets.



Thanks



Code & Paper & Data