



Zolly: Zoom Focal Length Correctly for Perspective-Distorted Human Mesh Reconstruction

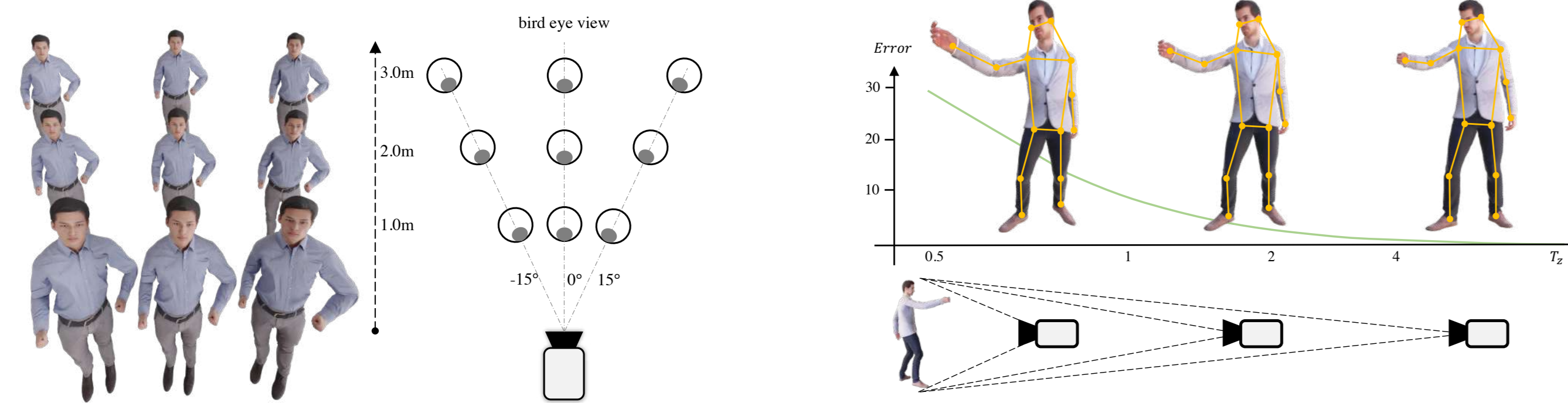
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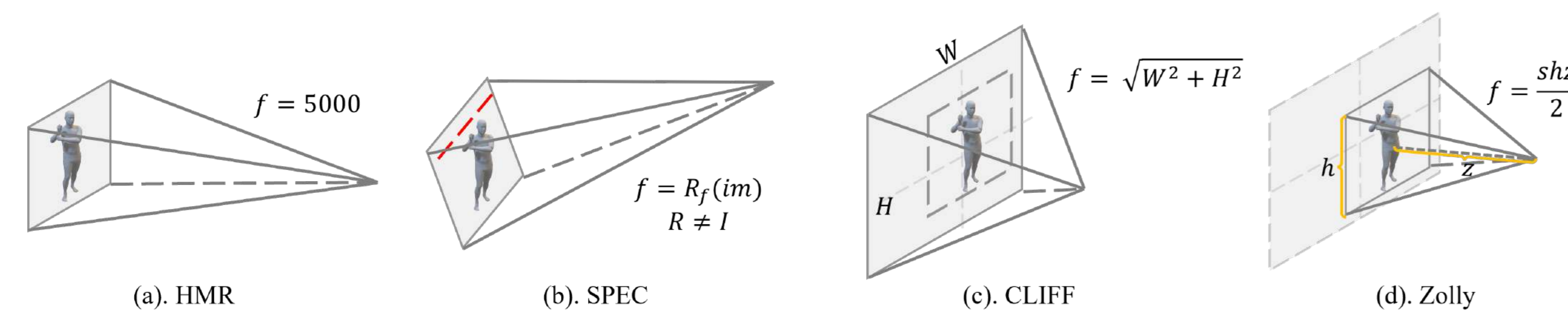
Introduction

- Motivation:** Existing 3D human mesh reconstruction methods use a constant f or estimate one based on the background context. Such f deviates a lot from distorted images caused by perspective projection. The distortion is directly caused by the distance and the facing angle to the camera center. Close-view shots could cause distortion on human bodies, which could be used to calculate the f .



- Contribution**
A novel camera system for the perspective-distorted 3DHMR task.
A new neural model, and a hybrid re-projection loss.
A new and the first synthetic dataset, PDHuman, for perspective-distorted 3D human pose estimation.

Camera System Design



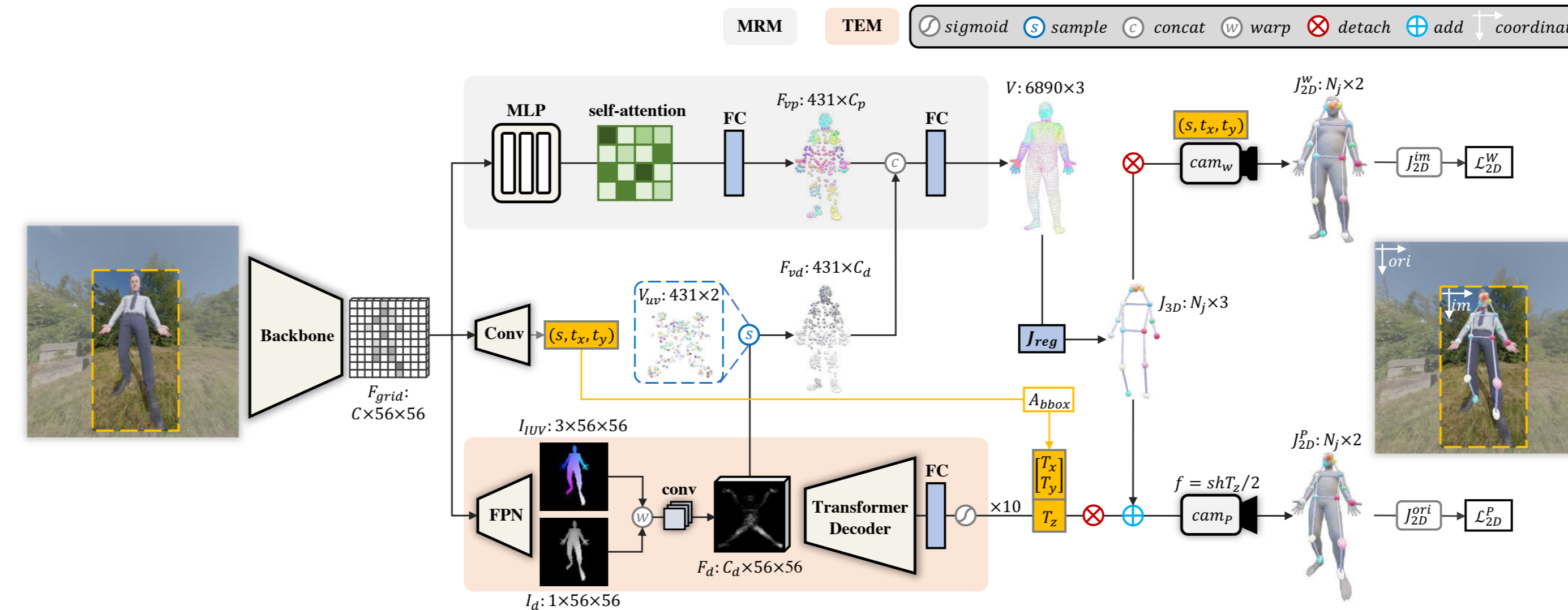
- HMR.** $f = 5000$. Most methods follow this setting.
- SPEC.** The f is estimated by a network pre-trained on other datasets.
- CLIFF.** Use the diagonal length as f if no ground truth f .
- Zolly.** $f = shT_z/2$. Where T_z is the z-axis distance.

The weak-perspective camera parameters (s, t_x, t_y) , which represent 2D orthographic transformation, could be used to approximate the projection:

$$\begin{bmatrix} f(x + T_z)/T_z \\ f(y + T_y)/T_z \end{bmatrix} = \begin{bmatrix} s(x + t_x) \\ s(y + t_y) \end{bmatrix}, s \times T_z = f, T_x = t_x, T_y = t_y. \quad (1)$$

In previous methods, they either use a constant or estimated f , and calculate distance by $T_z = f/s$. On contrary, we estimate T_z based on human body distortion clues and calculate f by $f = s \times T_z$. (NDC space)

Proposed Pipeline



- Translation Estimation Module.** We use a Transformer to regress the z-axis distance from the warped I_{UV} image and use sigmoid then $\times 10$ to restrict T_z less than 10m.
- Mesh Estimation Module.** We adopt an MLP structure to predict the coordinates of a coarse mesh of the body, then up-sample the mesh using two fully connected layers.
- Loss Functions.** The total loss function is the summation of mesh loss, translation loss, and hybrid re-projection loss.

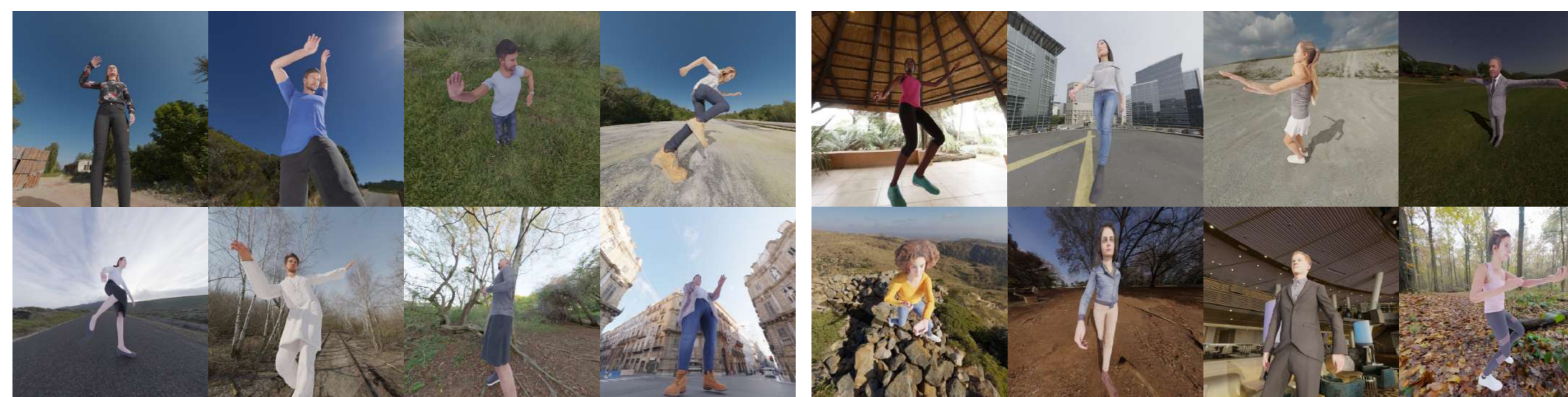
$$\mathcal{L}_{total} = \mathcal{L}_{Mesh} + \mathcal{L}_{Transl} + \mathcal{L}_{2D}^W + \mathcal{L}_{2D}^P \quad (2)$$

where the \mathcal{L}_{2D}^W is the weak-perspective and \mathcal{L}_{2D}^P is the perspective projection.

New Virtual Dataset: PDHuman

We propose the first dataset which aims for perspective-distorted 3D human pose estimation.

- Amount:** 126, 198 images in training and 27, 448 images in testing split.
- Annotations:** Camera intrinsic matrix, 2D/3D keypoints, SMPL parameters θ , β , and translation.
- Camera:** Use the dolly-zoom effect to generate random camera intrinsic matrices.
- Rendering:** Use human models from RenderPeople and body pose sequences from Mixamo, with HDRi images as backgrounds. Use Blender to render the RGB images.



Experiments

- Results on ordinary datasets and distorted datasets.

Method	Backbone	3DPW			Human3.6M	
		PA-JPE	MPJPE	PVE	PA-JPE	MPJPE
HMR	Res50	72.6	116.5	-	56.8	88.0
SPEC	Res50	52.7	96.4	-	-	-
CLIFF	HR48	43.0	69.0	81.2	32.7	47.1
Zolly	HR48	39.8	65.0	76.3	32.3	49.4

Method	Backbone	PDHuman (p5)			SPEC-MTP (p3)		
		PA-JPE	PVE	mIoU	PA-JPE	PVE	mIoU
HMR	Res50	62.5	106.7	21.7	73.9	145.6	16.0
SPEC	Res50	65.8	109.6	19.6	76.0	144.6	18.8
CLIFF	HR48	66.2	115.2	24.8	74.3	132.4	23.7
Zolly	HR48	49.9	82.0	26.5	67.4	126.7	30.4

Quality Results

Qualitative results of SOTA methods. Row 1: PDHuman test. Row 2, 3, 4: web images. Row 4: SPEC-MTP. Row 6: 3DPW. The number under each image represents predicted/ground-truth f , FoV angle, and T_z . The focal lengths here are all transformed to pixels in full image.

