

Gait Recognition Using 3-D Human Body Shape Inference

Paper Review 04/29

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- Title: Gait Recognition Using 3-D Human Body Shape Inference [1]
- Authors: Zhu, Haidong and Zheng, Zhaoheng and Nevatia, Ram
- Conference: IEEE/CVF Winter Conference on Applications of Computer Vision
- Year: 2023

Motivation

- Gait recognition based on 2-D images is difficult.
- Camera positions, carried objects, and clothing, etc.

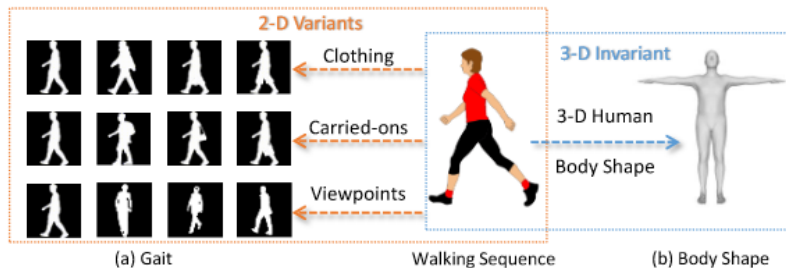


Figure 1: Variance of 2-D silhouette images

- 3-D body shape inference from silhouettes sequences
- Knowledge distillation for body shape

Proposed Method

Gait Feature Extraction

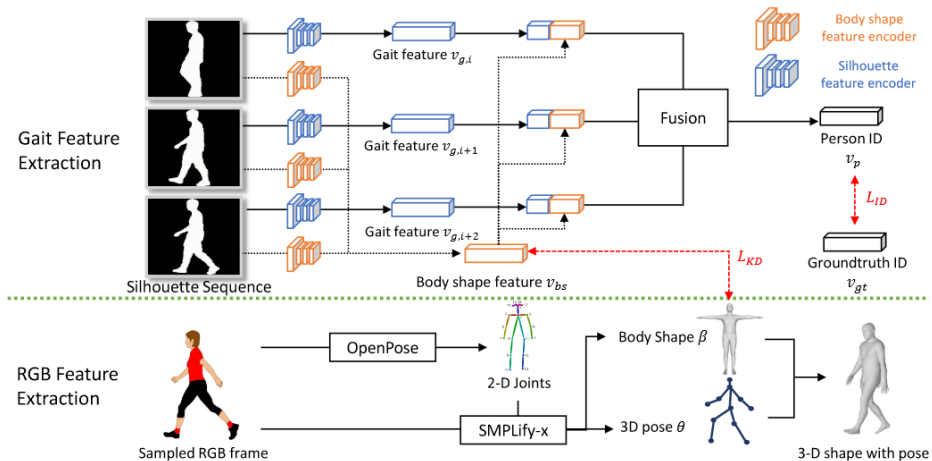


Figure 2: Overview of the proposed method

Proposed Method

Body Shape Feature Encoder

- sequence of silhouettes
- feature weights of different frame (12.5%)

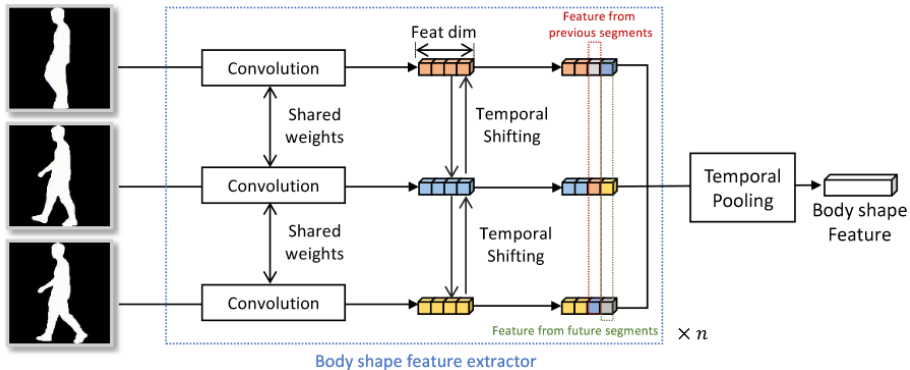


Figure 3: Body Shape Feature Encoder

Proposed Method

Body Prior Inference

- SMPLify-X
- Time consumption - select single RGB image

$$\begin{aligned} L_{KD} = & \mathbb{E}_{q(v_{br}, v_{bs} | C=1)} [\log h(v_{br}, v_{bs})] \\ & + \mathbb{E}_{q(v_{br}, v_{bs} | C=0)} [\log(1 - h(v_{br}, v_{bs}))] \\ h(s, t) = & \frac{\exp(f_1(v_{bs})^T \cdot f_2(v_{br}))}{\exp(f_1(v_{bs})^T \cdot f_2(v_{br})) + \frac{N}{M}} \end{aligned} \quad (1)$$

Figure 4: Contrastive Representation Distillation

- CASIA-B
 - 20% of sequence for data distillation
 - three scenario
 - normal walking
 - carrying bags
 - different clothing
- OUMVLP
 - lack of RGB video sequence
 - using shape encoder pre-trained on CASIA-B

Experiments

Results

Probe	Stats	HBS	Method				Avg. Change
			GaitSet [6]	GaitPart [10]	GLN [15]	GaitGL [23]	
NM #5-6	Mean (\uparrow)	X	95.6	96.4	96.5	97.3	+0.3
		✓	96.0	96.6	96.8	97.5	
		Δ	+0.3	+0.2	+0.3	+0.2	
	STD. (\downarrow)	X	3.4	2.8	2.8	1.7	-0.1
		✓	3.2	3.0	2.4	1.6	
		Δ	-0.2	+0.2	-0.4	-0.1	
BG #1-2	Mean (\uparrow)	X	90.1	91.2	93.1	94.4	+0.4
		✓	90.5	91.7	93.6	94.8	
		Δ	+0.4	+0.5	+0.5	+0.4	
	STD. (\downarrow)	X	4.6	4.2	3.3	2.7	-0.3
		✓	4.2	3.5	3.2	2.7	
		Δ	-0.4	-0.7	-0.1	0.0	
CL #1-2	Mean (\uparrow)	X	75.1	78.6	81.5	83.5	+1.1
		✓	76.0	79.7	83.2	84.1	
		Δ	+0.9	+1.1	+1.7	+0.6	
	STD. (\downarrow)	X	6.2	5.8	5.5	8.0	-0.4
		✓	5.9	5.6	5.5	7.0	
		Δ	-0.3	-0.2	0.0	-1.0	

Experiments

Results

Method	Camera Positions														Mean
	0°	15°	30°	45°	60°	75°	90°	180°	195°	210°	225°	240°	255°	270°	
GEINet [35]	23.2	38.1	48.0	51.8	47.5	48.1	43.8	27.3	37.9	46.8	49.9	45.9	45.7	41.0	42.5
GaitSet [6]	79.2	87.7	89.9	90.1	87.9	88.6	87.7	81.7	86.4	89.0	89.2	87.2	87.7	86.2	87.0
GaitPart [10]	82.8	89.2	90.9	91.0	89.7	89.9	89.3	85.1	87.7	90.0	90.1	89.0	89.0	88.1	88.7
GaitGL [23]	84.2	89.8	91.3	91.7	90.8	91.0	90.4	88.1	88.2	90.5	90.5	89.5	89.7	88.8	89.6
GaitSet-HBS	79.0	87.9	90.4	90.6	88.4	89.2	88.4	82.3	87.1	89.6	89.6	87.7	88.4	86.9	87.5
GaitPart-HBS	82.4	89.1	91.1	91.3	89.8	90.2	89.7	84.8	88.0	90.3	90.3	89.2	89.4	88.4	88.9
GaitGL-HBS	84.7	90.2	91.4	91.7	90.9	91.0	90.5	88.4	88.7	90.5	90.5	89.6	89.6	88.9	89.8

Figure 5: OUMVLP Results

Experiments

Results

- criteria
 - performance
 - viewpoints
 - appearance variance
- novel viewpoints

- Relay on exist model
- Quality of RGB video sequence
- Using skeleton for gait recognition?

- [1] Haidong Zhu, Zhaoheng Zheng, and Ram Nevatia. “Gait recognition using 3-d human body shape inference”. In: *Proceedings of the IEEE/CVF Winter Conference on Applications of Computer Vision*. 2023, pp. 909–918.

Thanks for Listening

Q & A