

## **GlocalNet: Class-aware Long-term Human Motion Synthesis**

#### **Motivation**

- Aim Synthesis of long-term (> 6000 ms) human motion skeleton sequences across a large variety of human activity classes (> 50) to aid human-centric video generation.
- **Applications** Augmented Reality, 3D character animations, pedestrian trajectory prediction...
- Challanges long-term temporal dependencies among poses, cyclic repetition across poses, bi-directional and multi-scale dependencies among poses, variable speed of actions, and a large as well as partially overlapping space of temporal pose variations across multiple class/types of human activities.



Overview of our two-stage framework, GlocalNet.

#### **GlocalNet Architecture**

- sparse motion.
- ■Loss function with Join (**L**<sub>ME</sub>) as:

$$L = (\lambda_1$$

$$L = (\lambda_{1} * L_{J}) + (\lambda_{2} * L_{MF})$$

$$L_{J} = \sum_{i=1}^{t} ||X[i] - \widehat{X}[i]||_{2} \qquad L_{MF} = \sum_{i=1}^{t-1} ||V[i] - \widehat{V}[i]||_{2}$$

$$\hat{V}_{i} = \widehat{X_{i+1}} - \widehat{X}_{i}$$



Authors: Neeraj Battan, Yudhik Agrawal, Sai Soorya Rao, Aman Goel, and Avinash Sharma Publication: Battan, Neeraj, et al. "GlocalNet: Class-aware Long-term Human Motion" Research Center Name: Centre for Visual Information Technology Synthesis." IEEE/CVF Winter Conference on Applications of Computer Vision. 2021.

# R&D SH WCASE 2021

**Technology, Social Impact** 

■ In the first stage, GloGen generates the sparse motion trajectory of an activity, followed by the second stage, LocGen, that predicts the dense poses from the generated

The t-SNE plot of GloGen embedding subspace along with the plot of selected motion trajectories where multiple samples for different classes are represented as color-coded 3D points.

### Results

Activity Labels		Input frames				Output frames											
throwing	舟	舟	舟	舟	介	净			1	1	*	*	×	×	×	\$	
sitting down	舟	舟	舟	舟	伊	伊	ি	ি	内	图	R	R	R	R	R	R	
putting hands togeth	ier 👖	介	舟	中	$\mathbf{P}$		$\frac{1}{2}$	$\frac{1}{2}$	Ŧ	Ŧ	中日	中日	中日	中日	中日	中日	
brushing hair	侟	侪	杏	杏	査	査	畜	合	本日	本日	首	古	首	首	首	Ť	
drink	舟	舟	索	索	索	索	索	索	宋	东	索	索	索	索	东	杏	
stand up	中	中	中	中	侪	侪	守	中	守	守	守	守	守	守	守	守	

Output of GloGen using different activity labels and initial poses.

Models	cross-vi	cross-subje			
	MMD <sub>avg</sub>	<b>MMD</b> <sub>seq</sub>	<b>MMD</b> <sub>avg</sub>	MI	
SkeletonVAE	1.079	1.205	0.992	1.1	
SkeletonGAN	0.999	1.311	0.698	0.7	
c-SkeletonGAN	0.371	0.398	0.338	0.4	
SAGCN	0.316	0.335	0.285	0.2	
Ours (L <sub>J</sub> )	0.213	0.218	0.201	0.2	
Ours (L <sub>MF</sub> )	0.646	0.647	0.601	0.6	
Ours $(L_J + L_{MF})$	0.195	0.197	0.177	0.1	

Comparison of Our Method (GloGen) in terms of MMD on NTU RGB+D(2D)



A A 本日 官官官官 ct MD<sub>sea</sub>



