







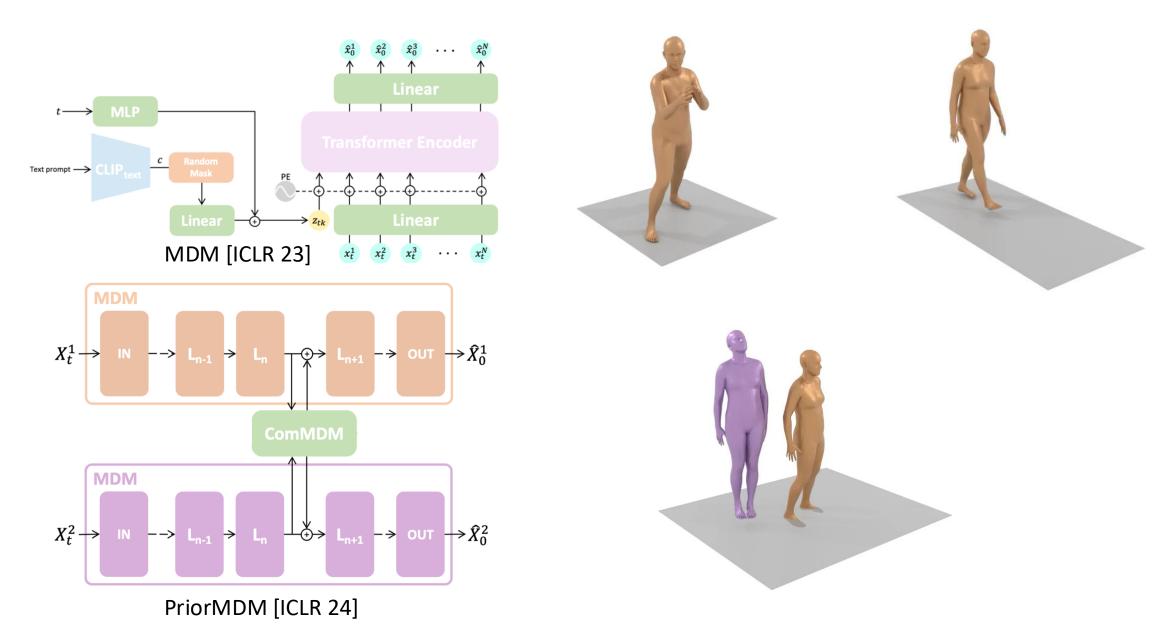
InterControl: Zero-shot Human Interaction Generation by Controlling Every Joint

NeurIPS 2024

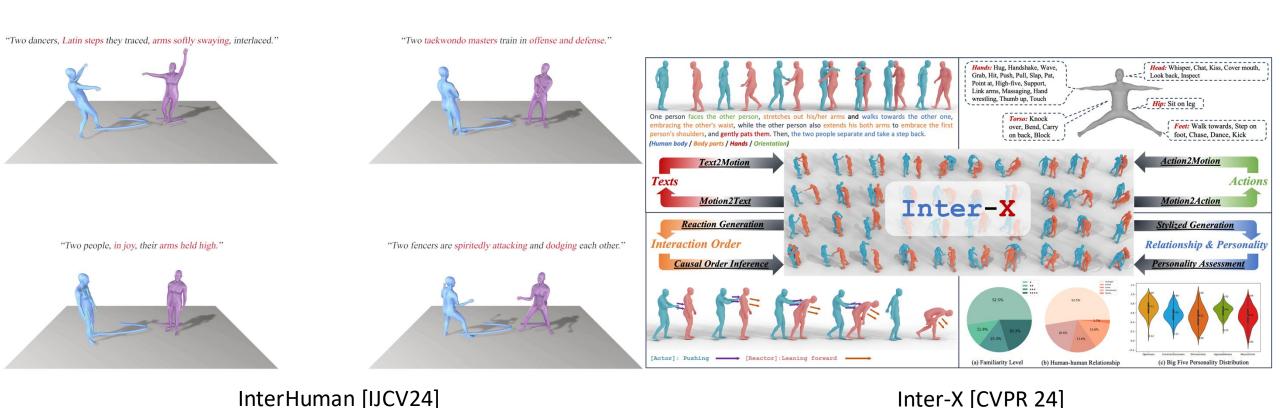
Zhenzhi Wang, Jingbo Wang, Yixuan Li, Dahua Lin, Bo Dai CUHK, Shanghai Al Lab, HKU

Code: https://github.com/zhenzhiwang/intercontrol

Human Motion Generation



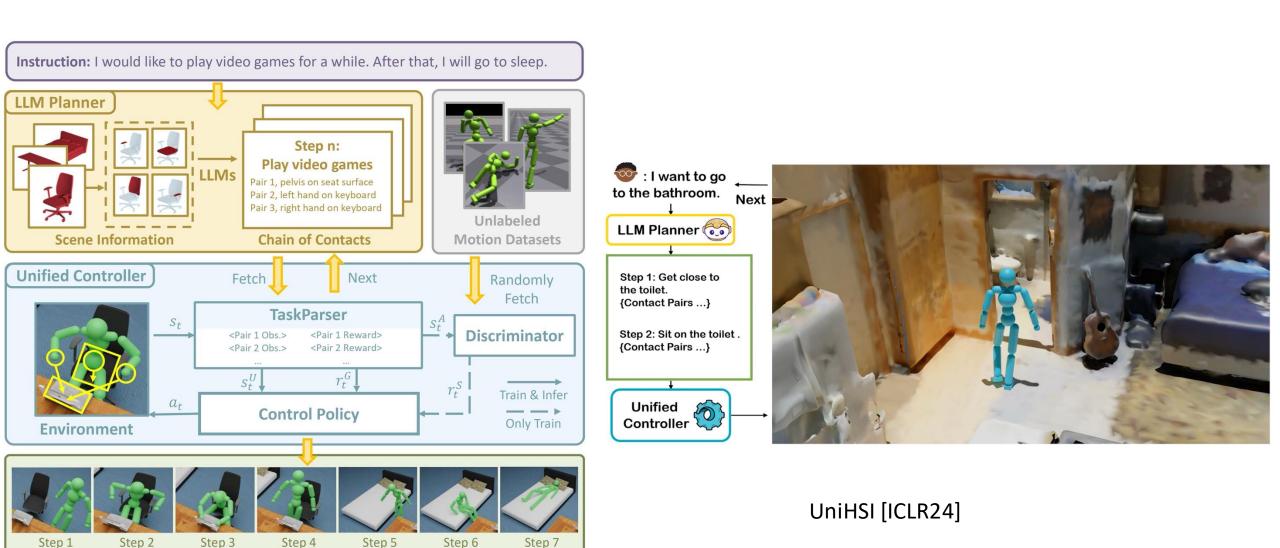
Multi-Person Interaction



Our work was done in Nov, 2023. Thus, back to that time, could we generate multi-person interactions only by single-person motion data?

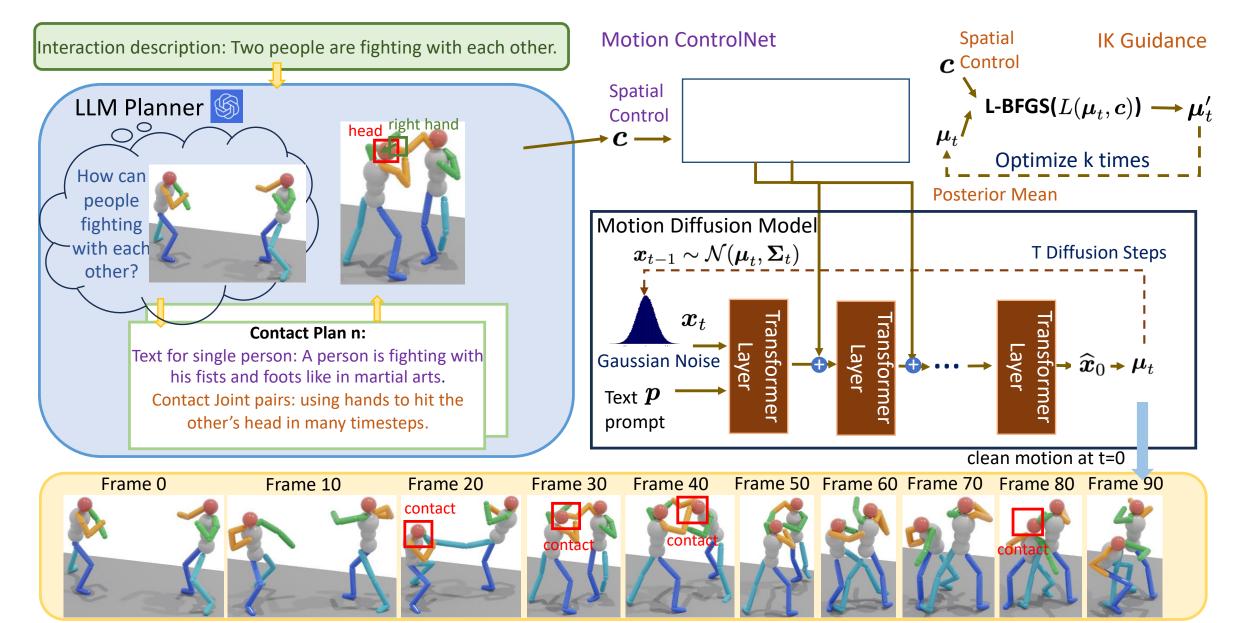
Advantages: the interaction will not be restricted by only two people defined by the dataset.

Definition of Interaction

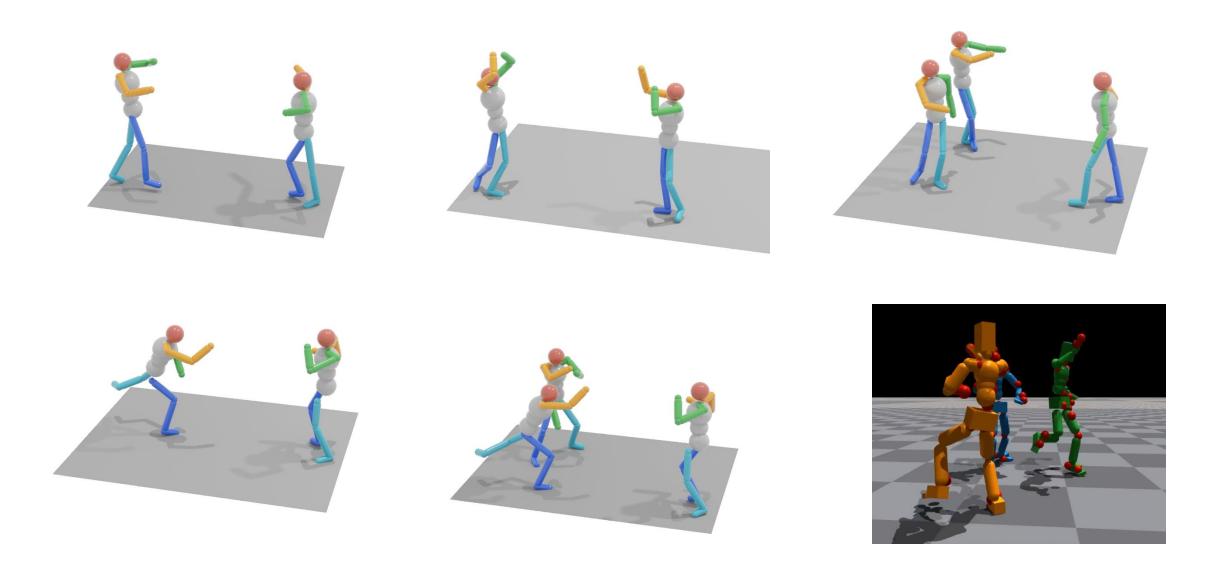


Yes! We find that interaction could be represented by joint distances. And the semantics of interactions could be understood by Large Language Models.

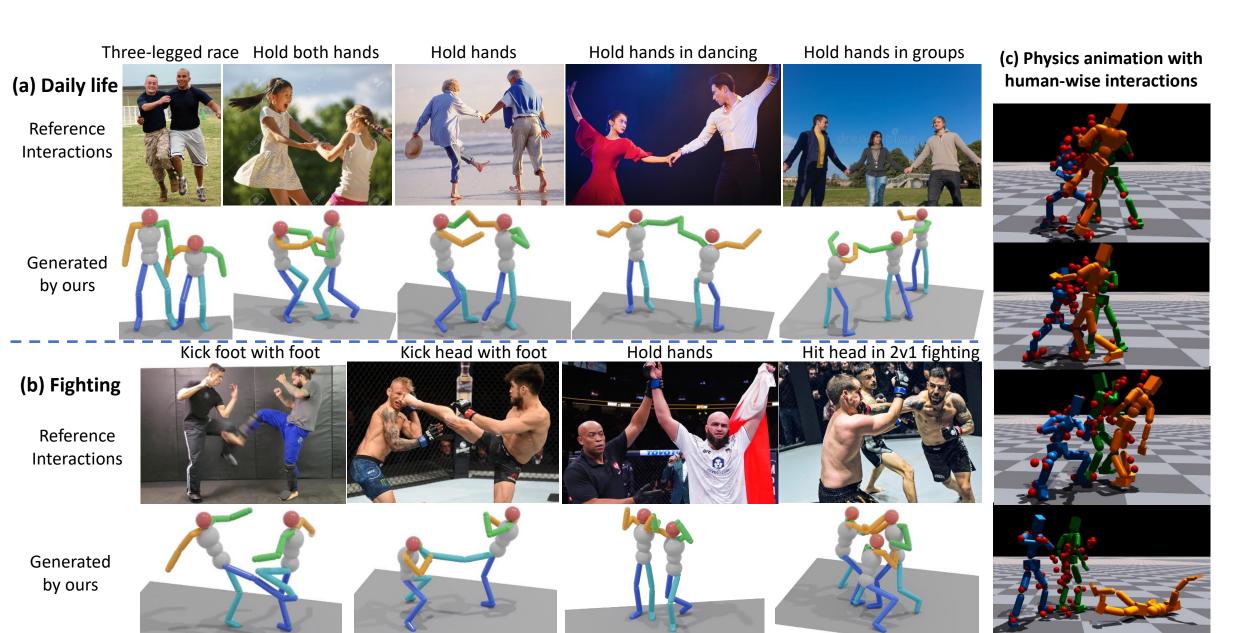
Spatial Control



Zero-shot Interaction Generation



Our interaction are realistic



Quantitative comparisons

Table 1: **Spatial control** results on HumanML3D [14]. \rightarrow means closer to real data is better. *Random One/Two/Three* reports the average performance over 1/2/3 randomly selected joints in evaluation. † means our evaluation on their model.

Method	Joint	FID ↓	R-precision ↑ (Top-3)	$Diversity \rightarrow$	Foot skating ratio ↓	Traj. err. ↓ (50 cm)	Loc. err. ↓ (50 cm)	Avg. err.↓ (m)
Real data MDM [55]	- No Control	0.002 0.544	0.797 0.611	9.503 9.446	0.0000 0.0943	0.0000 0.8909	0.0000 0.6015	0.0000 1.1843
PriorMDM [51] [†] GMD [27] [†] OmniControl [65] Ours	Root	0.498 0.276 0.218 0.159	0.586 0.655 0.687 0.671	9.167 9.245 9.422 9.482	0.0924 0.1108 0.0547 0.0729	0.3726 0.0987 0.0387 0.0132	0.2210 0.0356 0.0096 0.0004	0.4552 0.1457 0.0338 0.0496
OmniControl [65] Ours	Random one	0.310 0.178	0.693 0.669	9.502 9.498	0.0608 0.0968	0.0617 0.0403	0.0107 0.0031	0.0404 0.0741
Ours Ours	Random two Random three	0.184 0.199	0.670 0.673	9.410 9.352	0.0948 0.0930	0.0475 0.0487	0.0030 0.0026	0.0911 0.0969

Table 2: Evaluation on (left) spatial errors and (right) user preference in interactions.

Spatial Errors	Traj. err. (20 cm) \downarrow	Loc. err. (20 cm) \downarrow	Avg. err. (m) ↓
PriorMDM [51]	0.6931	0.3487	0.6723
Ours	0.0082	0.0005	0.0084

User-study	Preference		
PriorMDM [51]	18.8%		
Ours	81.2%		

Quantitative comparisons

Table 3: **Ablation studies** on the HumanML3D [14] dataset.

Item	Method	FID ↓	R-precision ↑ (Top-3)	$Diversity \to$	Foot skating ratio ↓	Traj. err. ↓ (50 cm)	Loc. err. ↓ (50 cm)	Avg. err.↓ (m)
(1)	Ours (random joint)	0.178	0.669	9.498	0.0968	0.0403	0.0031	0.0741
(2) (3)	w/o ControlNet w/ original c	0.965 0.227	0.621 0.656	9.216 9.544	0.1624 0.1004	0.0879 0.0697	0.0059 0.0042	0.1013 0.0785
(4) (5) (6)	w/o IK guidance IK guidance on x_0 w/ 1-st order grad	0.187 0.211 0.198	0.664 0.668 0.668	9.598 9.394 9.472	0.0704 0.1164 0.0987	0.8569 0.0907 0.0879	0.4553 0.0088 0.0096	0.6557 0.0981 0.0877
(7) (8)	sparsity = 0.25 $sparsity = 0.025$	0.248 0.255	0.671 0.663	9.442 9.520	0.0801 0.0705	0.0106 0.0015	0.0007 0.0001	0.0546 0.0067

Table 4: Inference time analysis on a NVIDIA A100 GPU.

Sub-Modules	MDM	+ Control Module	$\mid \textbf{+Guidance}\ t \in [10,999]$	$\mid \textbf{+ Guidance } t \in [0,9]$
Time (s)	39.1	57.3	76.5	80.1

Qualitative comparisons

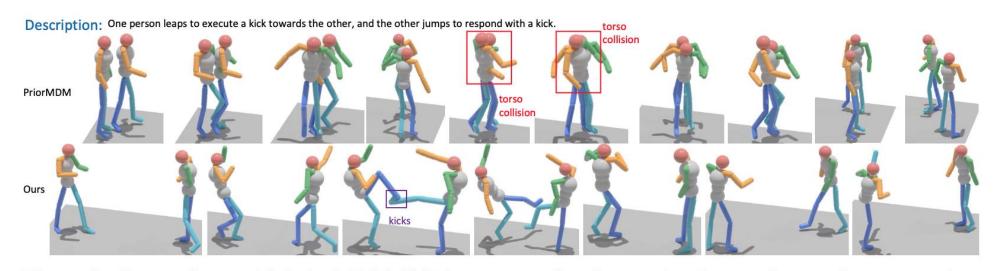


Figure 3: Comparison with PriorMDM [51] in user-study of zero-shot human interaction generation.

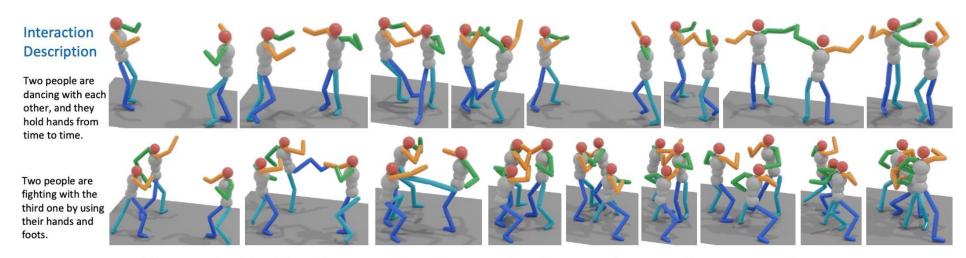


Figure 4: Qualitative results of zero-shot human interaction generation.

Thank you!