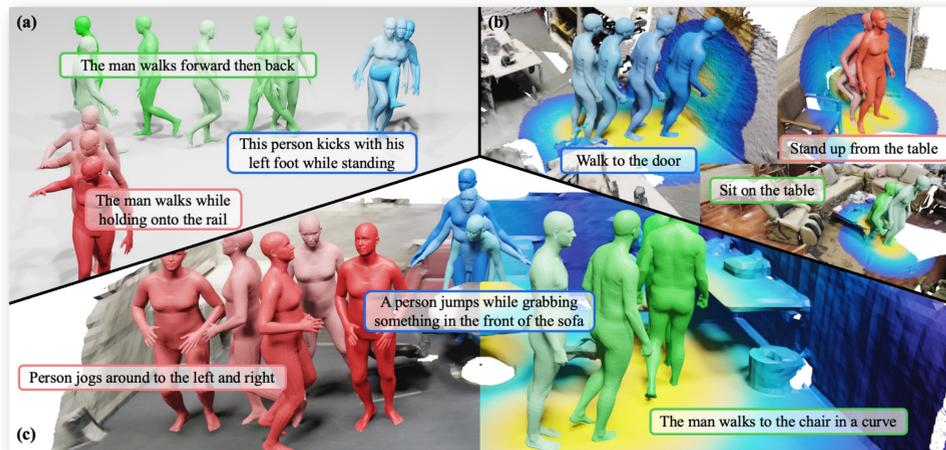




Introduction



We propose to **leverage scene affordance as an intermediate representation** to facilitate language-guided human motion generation in 3D scenes.

Challenges:

- The inherent complexity of marrying 3D scene grounding and conditional motion generation
 - ❖ Impede the model's ability to generalize to novel scenarios.
- The generative models' dependency on large volumes of high-quality paired data
 - ❖ Lack large-scale, motion-diverse, and semantic-rich HSI.

Contributions

- We introduce a novel two-stage model that **incorporates scene affordance as an intermediate representation**, facilitating language-guided human motion synthesis in 3D environments.
- We demonstrate our method's **superiority** over existing motion generation models on HumanML3D and HUMANISE benchmarks.
- Our model showcases remarkable **generalization capabilities**, performing impressively in generating human motions within **unseen** scenarios.

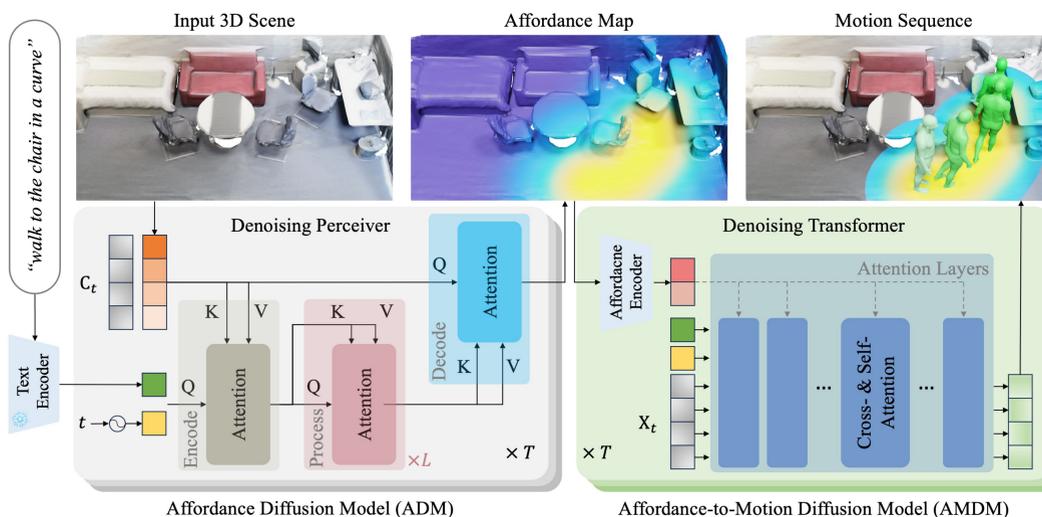
Method

Affordance Map

- We derive the scene affordance map from the distance field between scene points and human skeleton joints

$$c(n, j) = \exp\left(-\frac{1}{2} \frac{d(n, j)^2}{\sigma^2}\right) \quad \mathbf{C} = \text{max-pool}(c_1, c_2, \dots, c_F)$$

- ❖ Enhance the 3D grounding
- ❖ Provide a nuanced understanding of the geometric



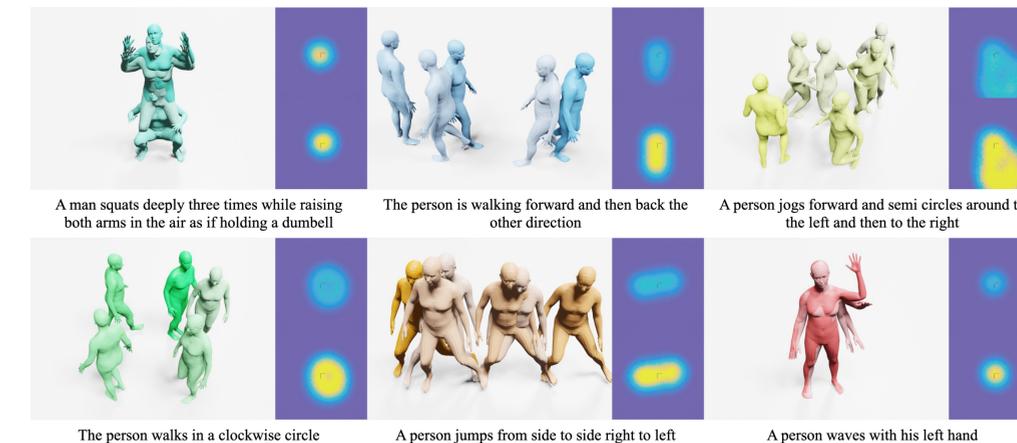
Results

Results on HumanML3D [Guo et al., CVPR 2022]

Quantative Results

Model	R-Precision ↑			FID ↓	MultiModal Dist. ↓	Diversity →	MultiModality ↑
	Top 1	Top 2	Top 3				
Real	0.511±.003	0.703±.003	0.797±.002	0.002±.000	2.974±.008	9.503±.065	-
Language2Pose [3]	0.246±.002	0.387±.002	0.486±.002	11.02±.046	5.296±.008	7.676±.058	-
T2M [29]	0.457±.002	0.639±.003	0.740±.003	1.067±.002	3.340±.008	9.188±.002	2.090±.083
MDM [76]	0.319±.005	0.498±.004	0.611±.007	0.544±.044	5.566±.027	9.559±.086	2.799±.072
Ours	0.341±.010	0.514±.016	0.625±.011	0.352±.109	5.455±.073	9.772±.117	2.835±.075
MDM [†] [76]	0.418±.005	0.604±.005	0.707±.004	0.489±.025	3.631±.023	9.449±.066	2.873±.111
Ours [†]	0.432±.007	0.629±.007	0.733±.006	0.352±.109	3.430±.061	9.825±.159	2.835±.075

Qualitative Results



Results on HUMANISE [Wang et al., NeurIPS 2022]

Quantative Results

Model	goal dist. ↓	APD ↑	contact ↑	non-collision ↑	quality score ↑	action score ↑
cVAE [84]	0.422±.011	4.094±.013	84.06±.716	99.77±.004	2.25 ± 1.26	3.66 ± 1.38
one-stage @ Enc	0.326±.013	5.510±.019	76.11±.684	99.71±.014	2.60 ± 1.24	3.88 ± 1.32
one-stage @ Dec	0.185±.014	4.063±.020	86.43±.845	99.76±.006	3.09 ± 1.34	4.18 ± 1.16
Ours @ Enc	0.156±.006	2.597±.008	95.86±.323	99.69±.007	3.46 ± 1.15	4.47 ± 0.84
Ours @ Dec	0.156±.006	2.459±.009	96.04±.298	99.70±.005	3.55 ± 1.19	4.44 ± 0.85

Qualitative Results



Results on Novel Evaluation Set



➤ Please refer to our project page for the animation videos and more results.