



## Both Ears Wide Open: Towards Language-Driven Spatial Audio Generation



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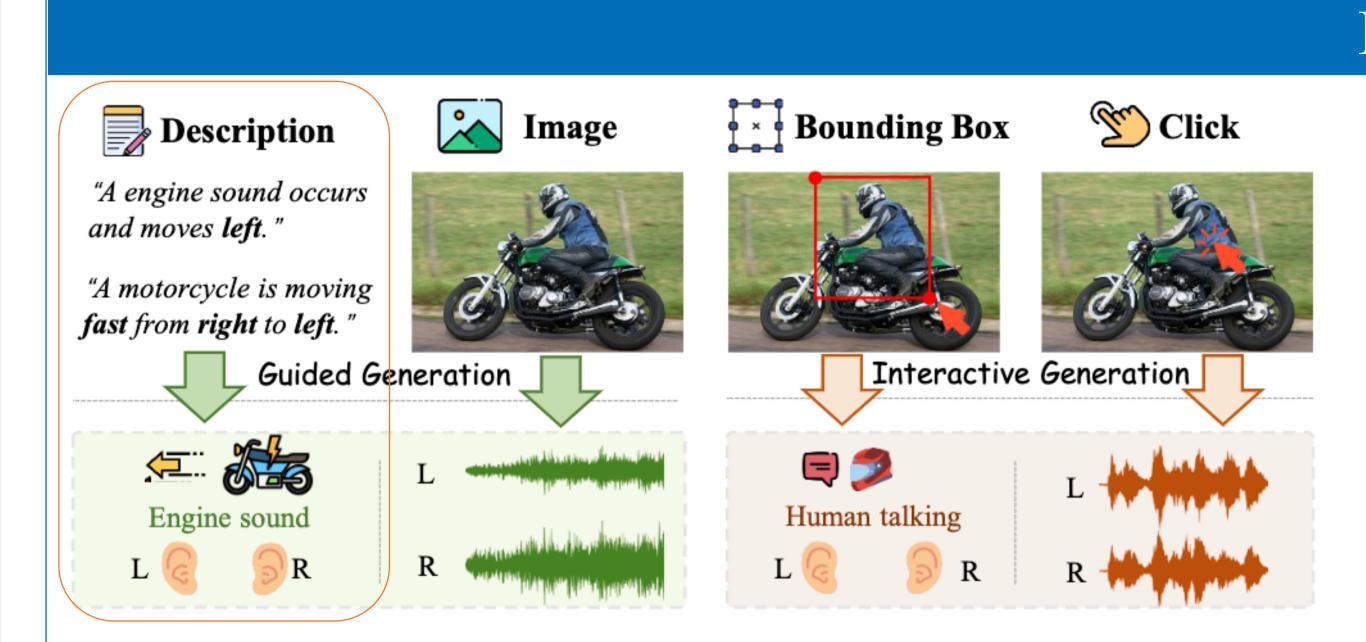
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Code



Demo Page



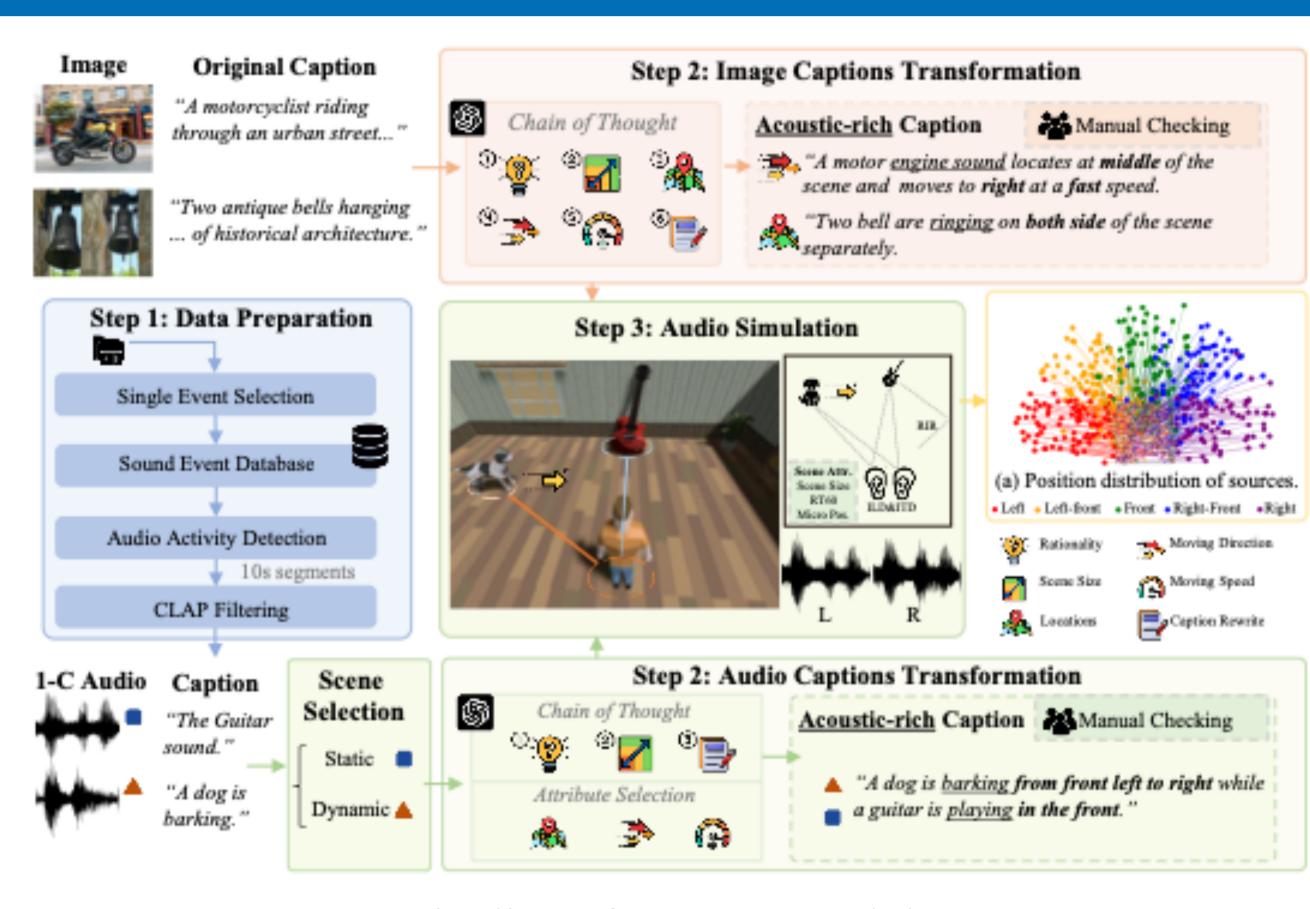
# Introduction

Why do this task?

Controlling stereo audio with spatial contexts to achieve immersive and attractive soundscapes that adheres physical world.

- To achieve this with following contribution
  - a) An open-source, large-scale, stereo audio dataset with spatial captions.
  - b) An one-stage, controllable, spatial audio generation framework.
  - c) a series of subjective and objective metrics based on ITD and opinion score.

#### Dataset: BEWO-1M



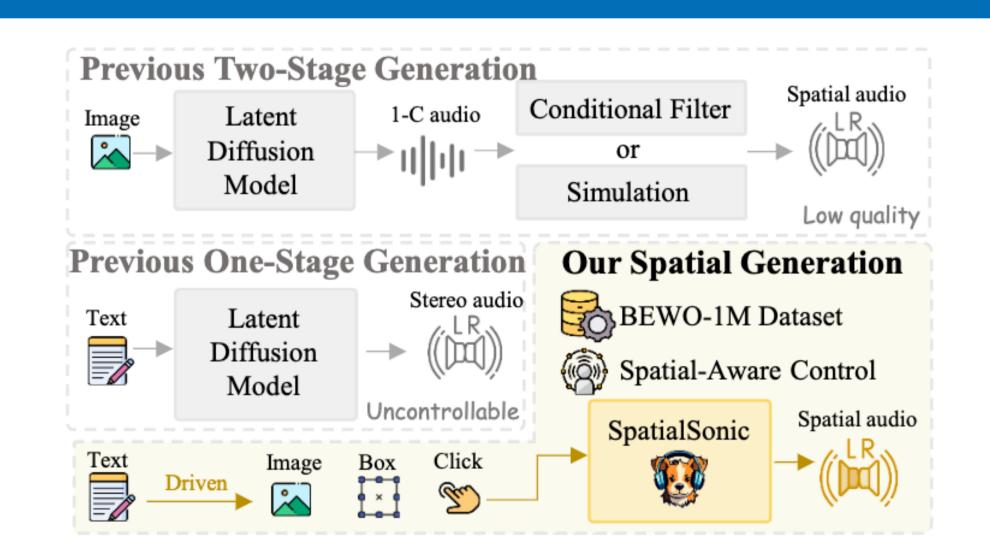
Tools	Deteget	Duration	Num. of	Paired
Task	Dataset	(hours)	Audios	Type
	LAION-Audio (Wu et al., 2023)	4.3k	633k	Text
	WavCaps (Mei et al., 2024)	7.5k	403k	Text
	AudioCaps (Kim et al., 2019)	110	46k	Text
Event	SoundDescs (Koepke et al., 2022)	1.1k	33k	Text
Event	Clotho (Drossos et al., 2020)	23	25k	Text
	Audio Caption (Wu et al., 2019)	10.3	3.7k	Text
	VGG-Sound Chen et al. (2020)	550	200k	Video
	AVE Tian et al. (2018)	11.5	4k	Video
	PicoAudio (Xie et al., 2024b)	15.6	5.6k	Text
Temporal	AudioTime (Xie et al., 2024a)	15.3	5.5k	Text
	CompA-order (Ghosh et al., 2024)	1.5	851	Text
	SimBinaural (Garg et al., 2023)	116	22k	Video
Spatial	FAIR-Play (Gao & Grauman, 2019)	5.2	1.9k	Video
	YT-ALL (Morgado et al., 2018)	113.1	1.1k	Video
	MUSIC (Zhao et al., 2018)	23	0.7k	Video
	BEWO-1M (Ours)	2.8k	1,016k	Text
	BEWO-1M (Ours)	54	2.3k	Image

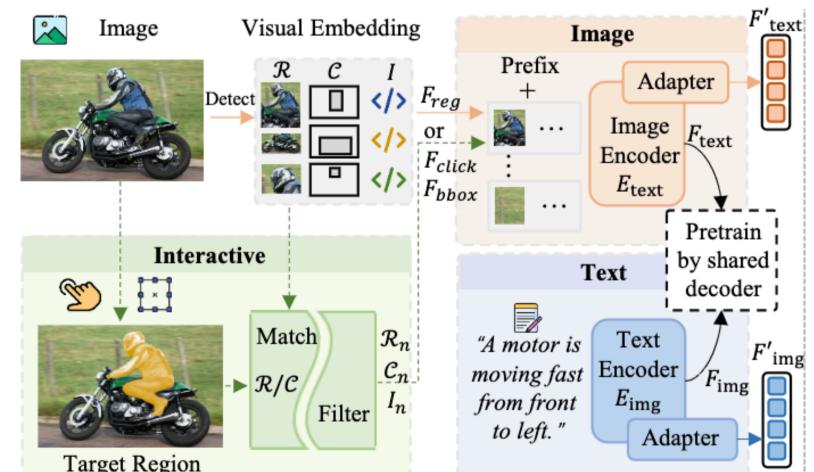
Pipeline of our proposed dataset.

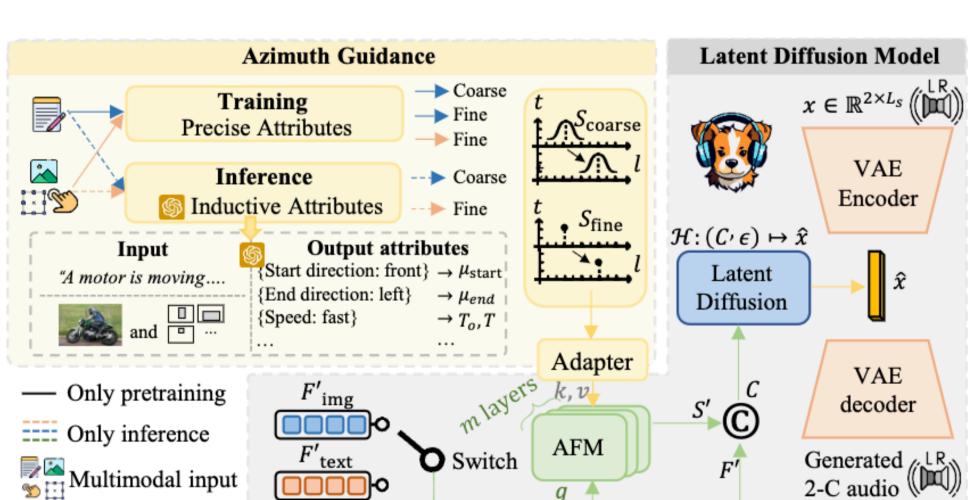
Comparison of datasets.

- > Developing a semi-automated pipeline to create an open-source, large-scale, stereo audio dataset with spatial captions, BEWO-1M and supporting both large-scale training and precise evaluation.
- Totally, we constructed 2.8k hours of training audio with more than 1M audio-text pairs and approximately 17 hours of validation data with 6.2k pairs. To the best of our knowledge, this work represents the first attempt to address these issues.

### Model: SpatialSonic





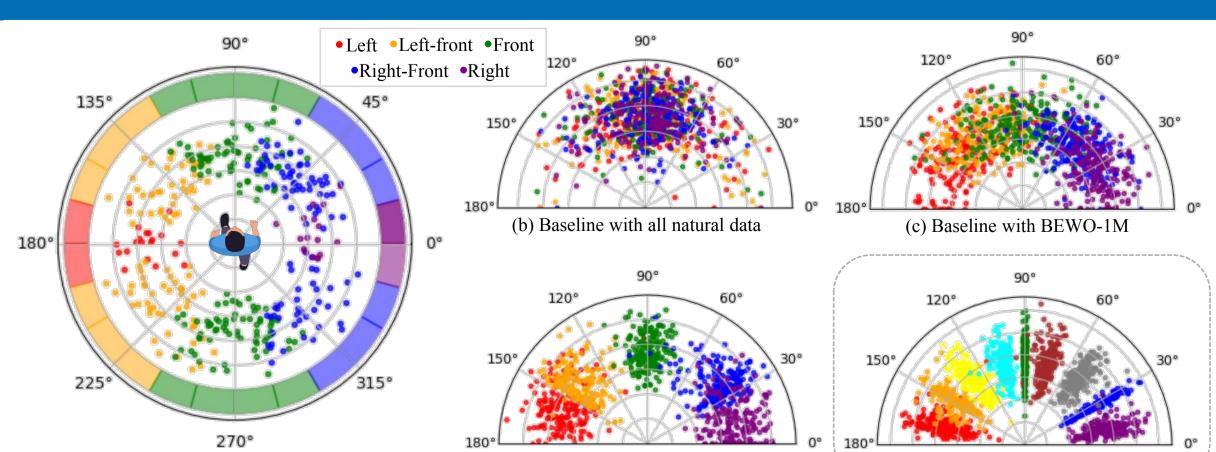


One stage generation model

Spatial audio generation pipeline based on DiT

> The proposed SpatialSonic takes azimuth as precise guidance and generate spatial audio with one stage behaviour.

### **Evaluation: ITD and MOS**



(d) Our coarse guidance with BEWO-1M

(a) Truman i erception	(a) Our course guidance with DD WO 11V1	(1)
Visualization of the	angle of generated	samples

Task	Method	Objective			Subjective	
		GCC MSE↓	CRW MSE ↓	FSAD↓	MOS-Events ↑	MOS-Direction ↑
Simulation	Simulation	-	-	-	4.94	4.95
T2A (SS-set)	AudioLDM2 <sup>†</sup>	46.59	50.17	1.61	3.57	3.53
	Make-An-Audio2†	38.83	43.12	0.97	3.58	3.59
	Stable-audio-open	38.73	34.36	0.63	3.73	3.76
	SpatialSonic(Ours)	27.20	15.86	0.17	3.78	3.84
T2A (SD-set)	AudioLDM2 <sup>†</sup>	45.08	42.88	0.94	3.37	3.34
	Make-An-Audio2†	48.55	47.88	1.09	3.38	3.30
	Stable-audio-open	45.76	48.60	0.53	3.68	3.58
	SpatialSonic(Ours)	44.36	31.91	0.26	3.86	3.71
T2A (DS-set)	AudioLDM2 <sup>†</sup>	38.96	50.96	2.48	3.29	2.97
	Make-An-Audio2†	35.37	48.54	2.11	3.24	3.31
	Stable-audio-open	32.63	36.30	0.87	3.60	3.61
	SpatialSonic(Ours)	22.51	13.75	0.31	3.80	3.83

Some important results of the subset (3/5)