

CLIPSep: Learning Text-queried Sound Separation with Noisy Unlabeled Videos

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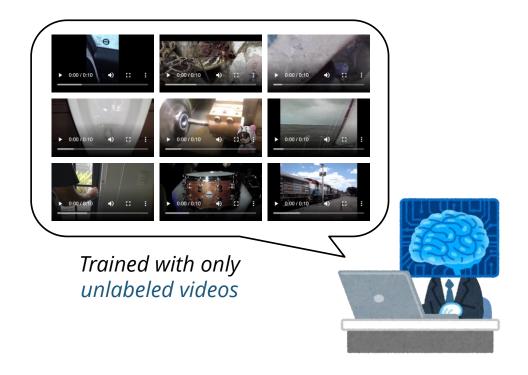


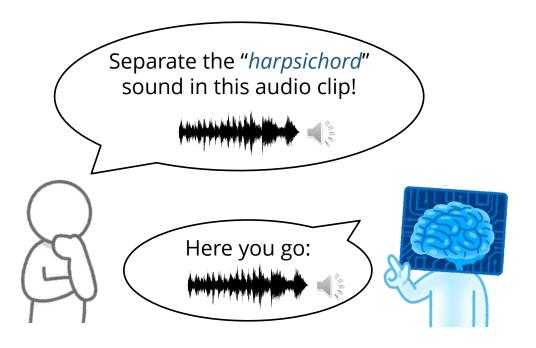
SONY

UC San Diego

Self-supervised Text-queried Sound Separation

Training Inference





Scalable to larger dataset

Natural text query-based interface

Data

MUSIC

(Zhao et al., 2018)



Violin



Acoustic guitar



Accordion

VGGSound

(Chen et al., 2020)



Hedge trimmer running

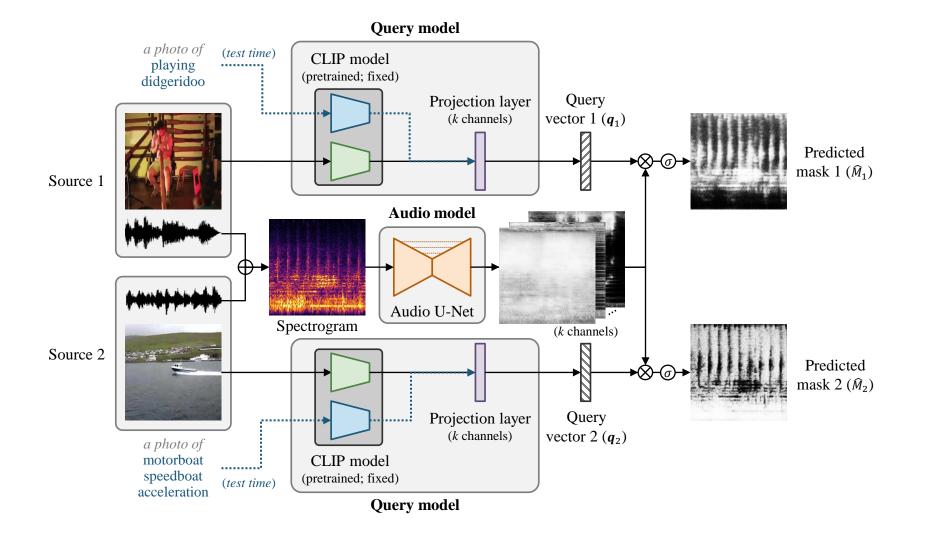


Dog bow-wow



Bird chirping, tweeting

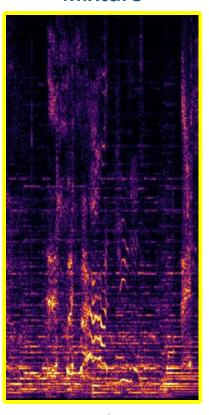
CLIPSep



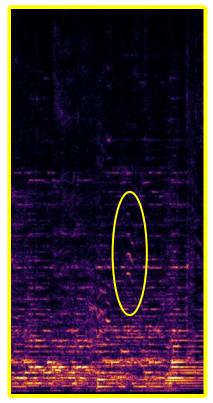
Demo – CLIPSep

Query: "playing harpsichord"

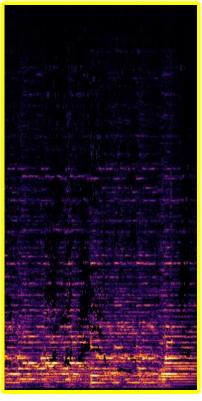
Mixture



CLIPSep

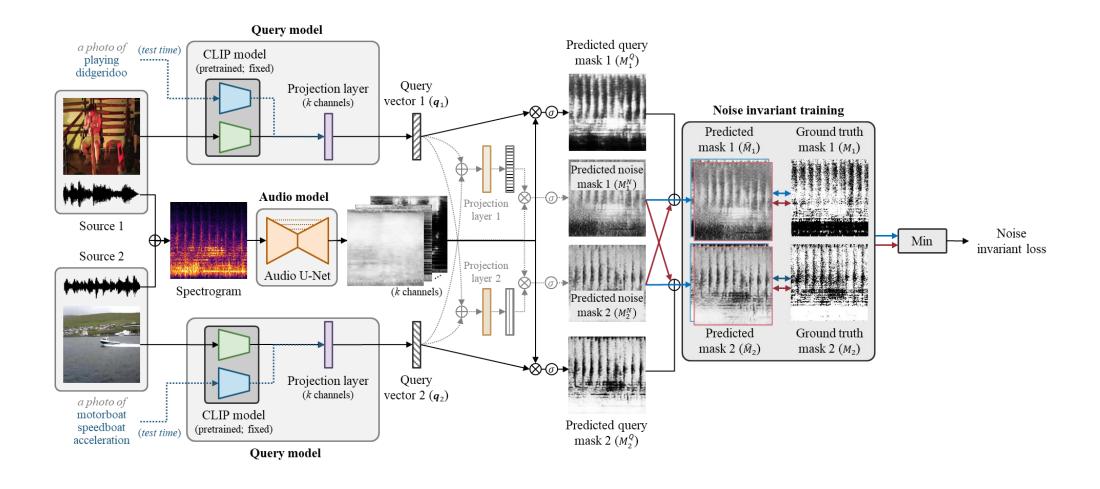


Ground truth





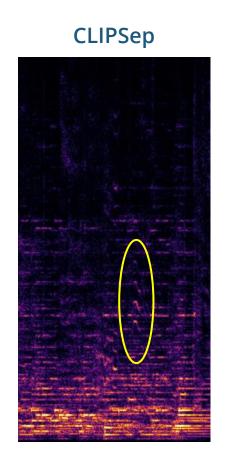
Noise Invariant Training (NIT)

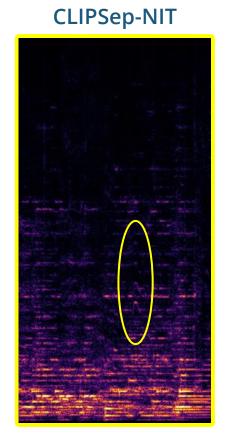


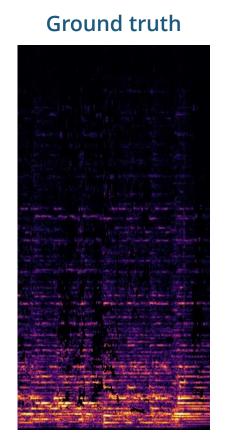
Demo – CLIPSep-NIT

Query: "playing harpsichord"

Mixture













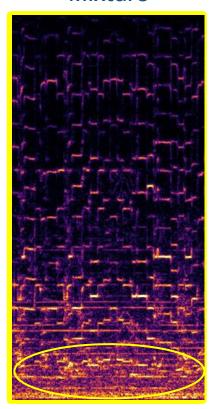
Quantitative Results

			MUSIC ⁺		VGGSound-Clean+	
Model	Unlabeled data	Post-proc. free	Mean SDR	Median SDR	Mean SDR	Median SDR
Mixture	-	-	4.49 ± 1.41	2.04	-0.77 ± 1.31	-0.84
Text-queried models						
CLIPSep	✓	✓	9.71 ± 1.21	8.73	2.76 ± 1.00	3.95
CLIPSep-NIT	✓	✓	$\textbf{10.27} \pm \textbf{1.04}$	10.02	$\textbf{3.05} \pm \textbf{0.73}$	3.26
BERTSep		√	4.67 ± 0.44	4.41	5.09 ± 0.80	5.49
CLIPSep-Text		✓	10.73 ± 0.99	9.93	5.49 ± 0.82	5.06

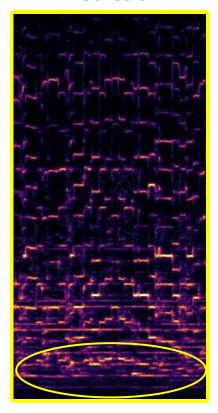
Demo - Noise Removal

Query: "playing bagpipe"

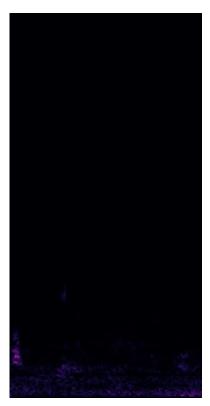
Mixture



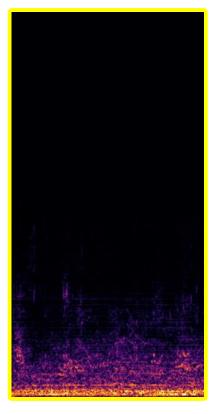
Prediction



Noise head 1



Noise head 2











Summary

CLIPSep

First text-queried universal sound separation model that can be trained using only unlabeled videos

Noise Invariant Training

An approach for training a querybased sound separation model with noisy data in the wild











Paper: <u>arxiv.org/abs/2212.07065</u> Demo: <u>sony.github.io/CLIPSep/</u> Code: <u>github.com/sony/CLIPSep</u>