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MUSIC SOURCE SEPARATION IN NOISY BRAZILIAN CHORO RECORDINGS

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1. MOTIVATION

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- Choro is considered the first musical style to originate in Brazil, dating back to the 1870s;
- Some historical recordings from the early 20th century include noise inherent to the process of recording and playing shellac records;

3.EXPERIMENTS

- The experiments were conducted to test the separation models on simulated historical choro recordings. To do that, 20 tracks were carefully selected from the test set of the choro dataset to ensure they were free of leakage;
- This work investigate the instrument separation task applied to historical recordings of this Brazilian music genre, using models originally trained on clean tracks.

2. BACKGROUND

- In recent years, deep learning Number approaches have achieved stateof-the-art performance in music source separation;
- chose to We base the investigation of this work on the Hybrid Demucs model;
- The model has been trained (from scratch) on a choro music data set containing 10 albums for training and 2 albums for validation.

er	Songbook	# Songs
	Altamiro Carrilho	13
	Benedicto Lacerda	12
	Chiquinha Gonzaga	12
	Choro Meets Bach	14
	Ernesto Nazaré 1	11
	Ernesto Nazaré 2	11
	Ernesto Nazaré 3	11
	Inéditos	14
	Jacob do Bandolim 1	12
	Jacob do Bandolim 2	12
	Pixinguinha	12
	Roda de Choro	12
	Severino Araújo	12
	Waldir Azevedo 1	12
	Waldir Azevedo 2	12
	Zequinha da Abreu	12

13

To simulate historical 78 RPM noisy recordings, we combine the clean tracks of choro test set and the 5 pre conditioned noise tracks according to:

 $C_{\text{noisy}} = \beta (G.N_{pp} + C_{\text{clean}})$

where β represents the global track gain, G is the gain applied to the pre-processed noise track N_{pp} to produce the desired SNR, and C_{clean} is the track extracted from choro test set.

- Noise gains were adjusted to induce SNR values of 10, 20, 30, and 40 dB for each track. In total, 400 noisy mixtures were generated according to this scheme (Figure Section 4.)
- To assess the quality of separation, three objective metrics are widely adopted in the literature: signal-to distortion ratio (SDR), signal-to-interference ratio (SIR) and signal-to-artifact ratio (SAR) (Tables Section 4).

4. RESULTS







Conclusions

-10 dB

-20 dB

-30 dB

-40 dB

-50 dB

-60 dB

70 dB

-10 dB

-20 dB

-30 dB

-40 dB

-50 dB

- -60 dB

-70 dB

The results demonstrate that the system is robust when dealing with tracks containing additive noise, even though it has been pre trained on clean recordings. Some families, such as percussion, face greater challenges in separation at lower SNRs, while others, such as wind instruments, show good results across all SNR levels. Some separation results, as well as the available for https://www02.smt.ufrj. noises used, listening at are br/~pedro.donadio/index_Lamir.html.



Average value of SDR, SIR and SAR computed for the 20 noisy tracks in test set with an SNR of 10 dB

> Listen to the noises and results here

