

# Polyphonic Avian Vocalizations Classification using Transfer Learning

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**Abstract**—In the application of audio classification, audio signals have various noises overlapping the signal of interest. The signal may also have more than one sound of interest embedded. Thus, classification models of a single sound are, at times, unrealistic for application. Thus, polyphonic classification models can be used to determine if multiple sounds of interest are present within a given audio source. In this work, we propose a new polyphonic variant of the BirdCLEF 2021 data set. The mixing of avian vocalizations is dependent on regional relationships. We propose a novel architecture to create mixed audio embeddings that can be used as initial features to perform transfer learning on polyphonic classification methods in general. We show that through experimentation, our proposed mixed audio embedding method enhances the performance on the task of multi-classification on our proposed polyphonic variant of BirdCLEF 2021.

**Index Terms**—BirdCLEF 2021

## I. To Do

### A. Rick

- 1) Make mixed dataset
- 2) Train mixed embedder
- 3) Write up in Experiment and Results about how A Framework for the Robust Evaluation of Sound Event Detection (<https://arxiv.org/abs/1910.08440>) works and set up code examples with ([https://github.com/audioanalytic/psds\\_eval](https://github.com/audioanalytic/psds_eval))
- 4) Write lit review
- 5) Pick a conference and get that info
- 6) Learn everything on here (<https://www.birds.cornell.edu/home/>)

### B. Mia

- 1) one

## II. INTRODUCTION

Why do we need to classify avian vocalizations, why does it need to be polyphonic, what is the BirdCLEF 2021 challenge and dataset, we perform this task on the edge?

## III. LITERATURE REVIEW

[1] proposed a way to address BirdCLEF 2021. Pick literature from the reference section of [1] on other papers about BirdCLEF 2021. Refer to work from <https://www.birds.cornell.edu/home/> about why bird call classification matters. Refer to work from A Framework for the Robust Evaluation of Sound Event Detection (<https://arxiv.org/abs/1910.08440>) why polyphonic classification matters.

#### IV. DATA

Our polyphonic variant of BirdCLEF 2021 was created using the tool Scraper [2].

#### V. METHODOLOGY

##### A. Polyphonic Transfer Learning Embedding

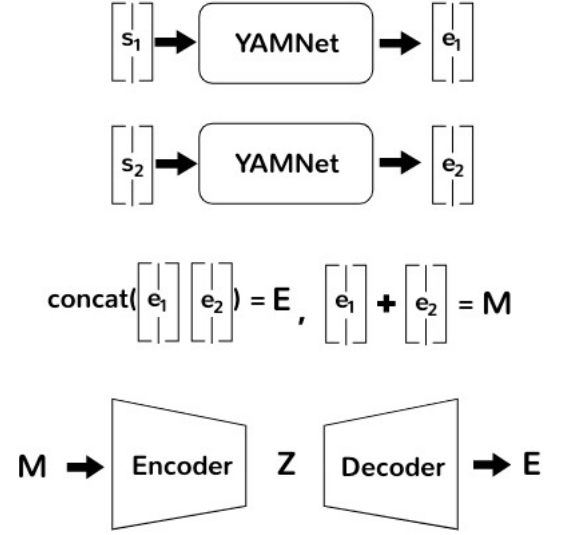


Fig. 1. Polyphonic Transfer Learning Embedding

## VI. EXPERIMENT AND RESULTS

To score our methodology on our proposed data set, we used the proposed framework by [3]. This consist of the following metrics: metricA, metricB, etc.

*A. Summarize the work of [3]*

## VII. CONCLUSION

### ACKNOWLEDGMENTS

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## REFERENCES

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