# Sounds of Nature

Semester Project Presentation 09/03/2023

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### Uncovering the Secrets of Nature



Ecologist experts record the bioacoustic activity for months

# Recognition of species and groups of species species 91% 14% 9% Soundscapes Recognition and Environmental Analysis



### Synergizing Efforts: Structuring Effective Teamwork



Remote and asynchronous communication



Taking Control of Your Day with Self-Management



Agile Working Framework (1 week sprint)

# Outline

- Recognition of species and groups of species
- Soundscapes Recognition and Biodiversity Analysis

# Recognition of species and groups of species

### **General Context**

 In this project, artificial intelligence is used to solve an ecological problem using a system that recognizes species and groups of species in nature, especially the vocalizations of animals.





# Objective

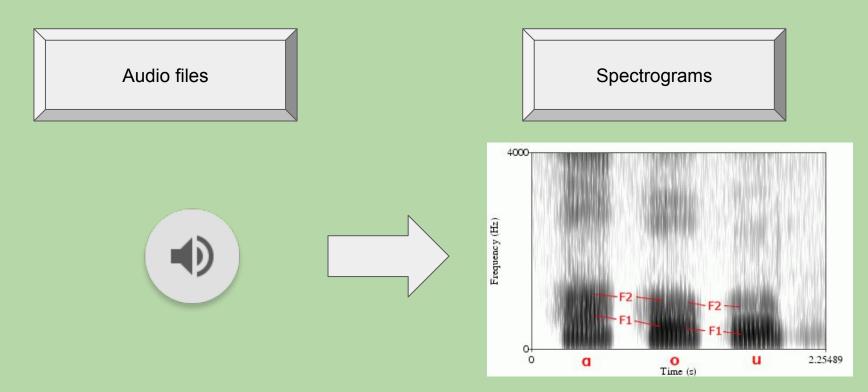
• The main goal is to develop a classifier capable of recognising animal species by their sounds:



Based on the animal's vocalizations, an AI algorithm can determine the species.



### Datasets



### Preprocessing

### Audio dataset



- Audio files < 5 seconds are repeated several times to reach 5 seconds
- Resampling (22,000 Hz)



# Audio converted into spectrograms





Linear spectrogram

Mel spectrogram

### Data Augmentation

### First Method

Solving the problem of unbalanced classes :

Adding the noise extracted from part 1 of the project to the samples of underrepresented classes

= > This method is applied before starting the training of the models and directly to the audio files.

### Second Method

Adding new samples to all the classes equally :

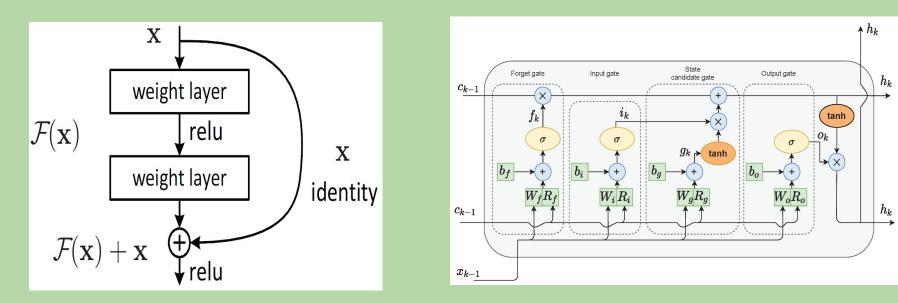
Using frequency shift augmentations on the spectrograms

= > This method is used during the training of the models and on the spectrograms.

### **R-CNN**

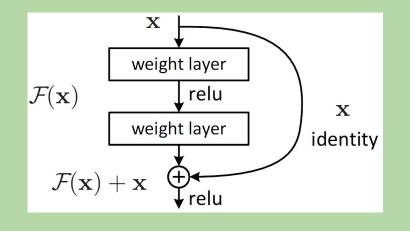
• **<u>Resnet</u>**: A block of Convolutions

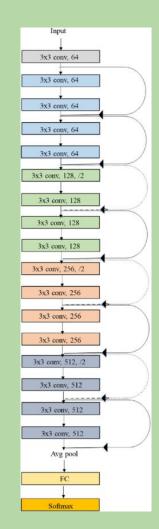
• **LSTM**: Memory cells



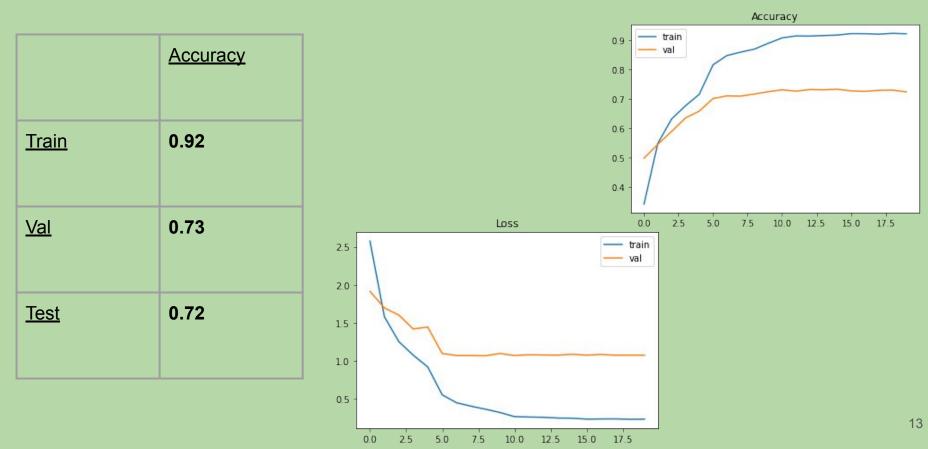
### ResNet18

• Number of parameters : 11,210,883

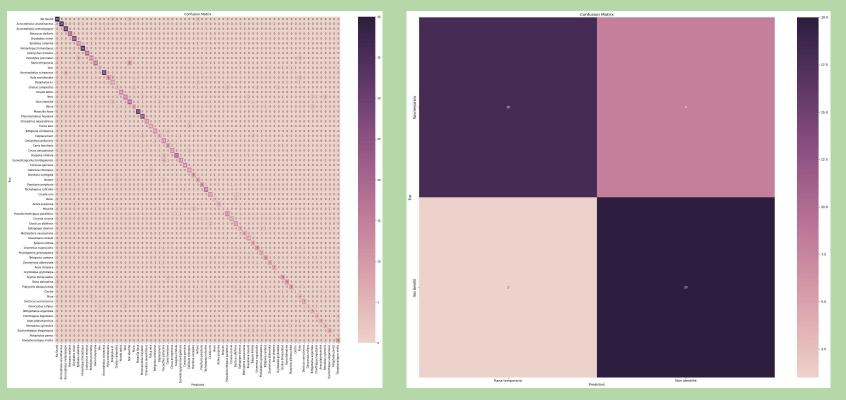




### The best Model : ResNet18



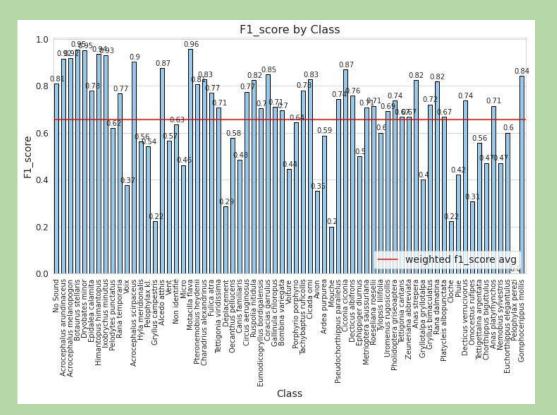
### **Confusion Matrix**



#### **Confusion matrix**

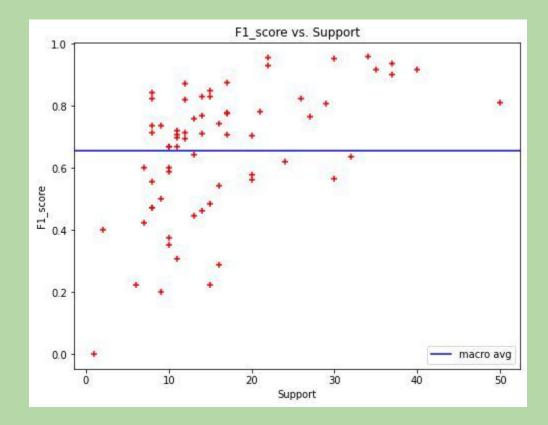
#### Relevant confusion matrix

### F1-score by class

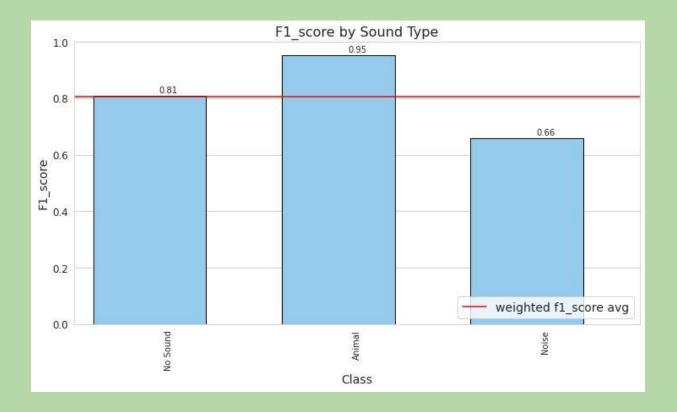


Class	F1-score (avg)
Animal	0.69
Noise	0.41

### F1-score vs support



# F1-score by Sound-Type



# Perspectives

What are the relevant research perspectives?

- Hierarchical loss.
- Updating the softmax function to handle audios that contains multiple species.

# Soundscapes Recognition and Environmental Analysis

## **Problem Statement**

#### Given a bioacoustic activity



#### Task n°1: Soundscapes Recognition

Find the labels associated with a bioacoustic activity:

Tree line evolution

- Blue: forest that has descended in altitude since the 50's
- Red: forest that has increased in altitude since the 50's

#### Sensor position

- Estive: 200+m above the upper tree line
- Edge: At the tree line

#### Task n°2: Environmental Analysis

Estimate ecosystem variables

#### Examples

- Distance to a road
- Average altitude
- Average temperature evolution

### Overcoming the Challenges of Handling a Large Dataset



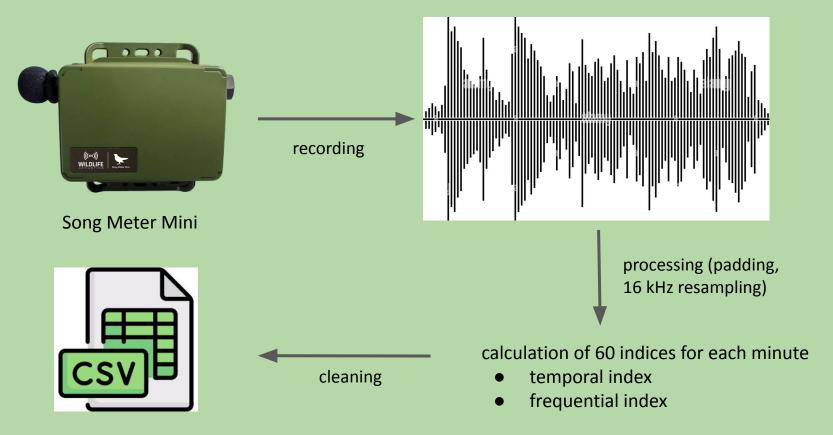
- 00:00 SPRING 08:00 16:00
- Months of bioacoustic activity recordings (1.1TB)
- **16 different geographical** sites (red pins)
- One day of the same season chosen for each site
- Three 30-minute samples throughout the day
- New representation of the database (~10GB)

### Index Approach



### **Examples**

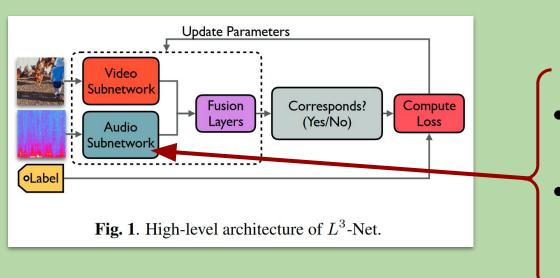
- ACI →high values = indicate storms, intermittent rain drops falling from vegetation
- ACI → low values = recordings with consistent cicada noise



3116 rows × 64 columns

### Approach with Embedded Indexes

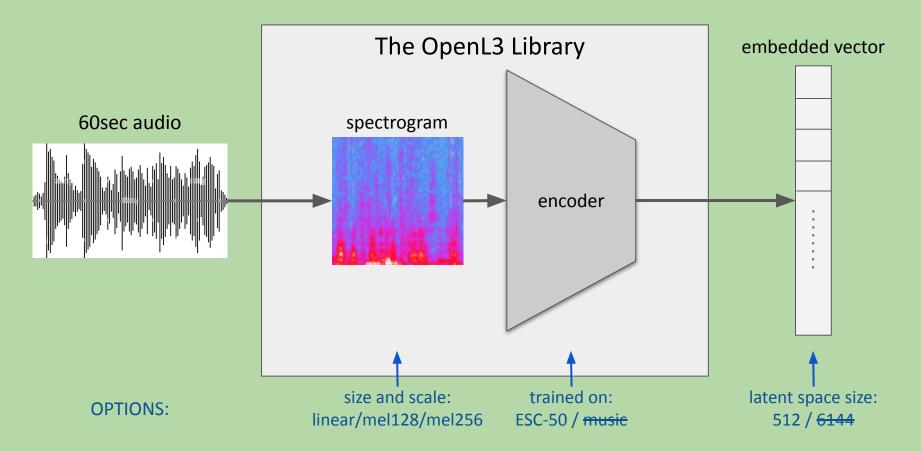
### The OpenL3 Library



- contains an encoder that is trained
- the encoder summarizes audio (as spectrograms) into a vector of smaller dimension

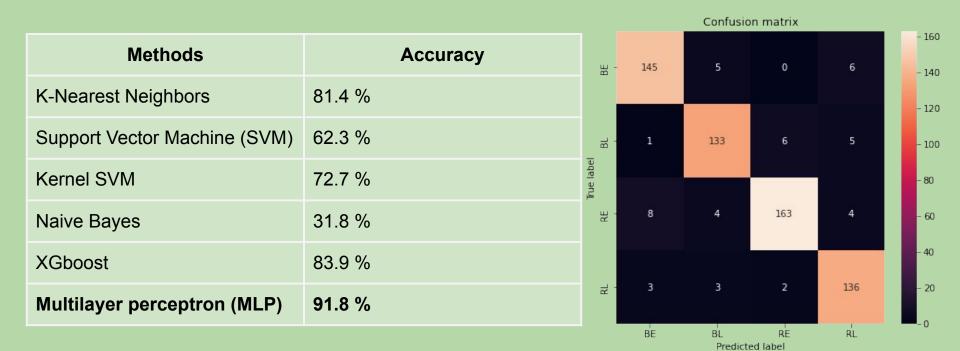
**Source**: Look, Listen and Learn More: Design Choices for Deep Audio Embeddings Jason Cramer, Ho-Hsiang Wu, Justin Salamon, and Juan Pablo Bello. IEEE Int. Conf. on Acoustics, Speech and Signal Processing (ICASSP), pages 3852-3856, Brighton, UK, May 2019.

### Approach with Embedded Indexes

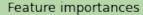


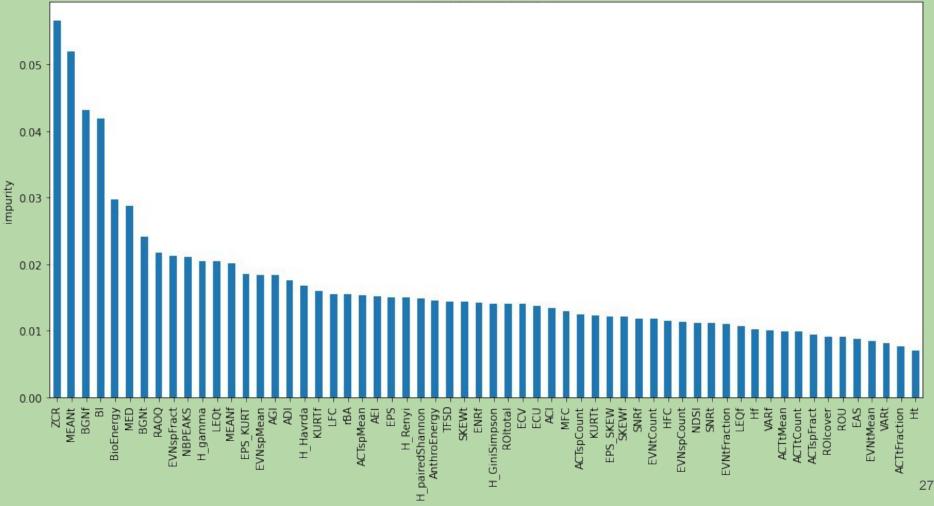
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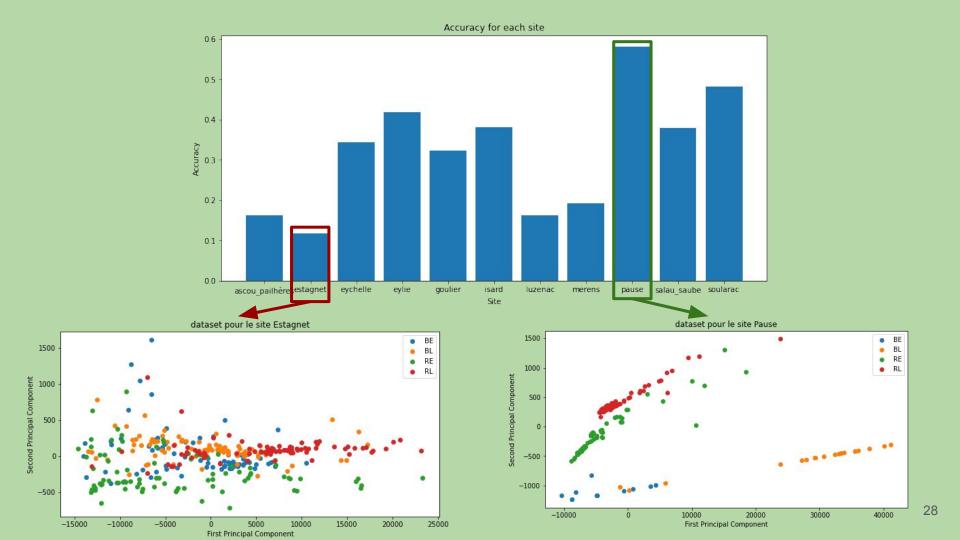
### Analysis of index approach



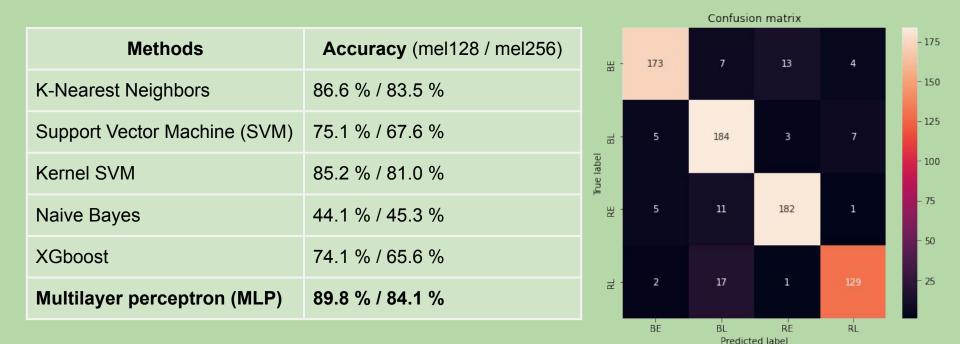
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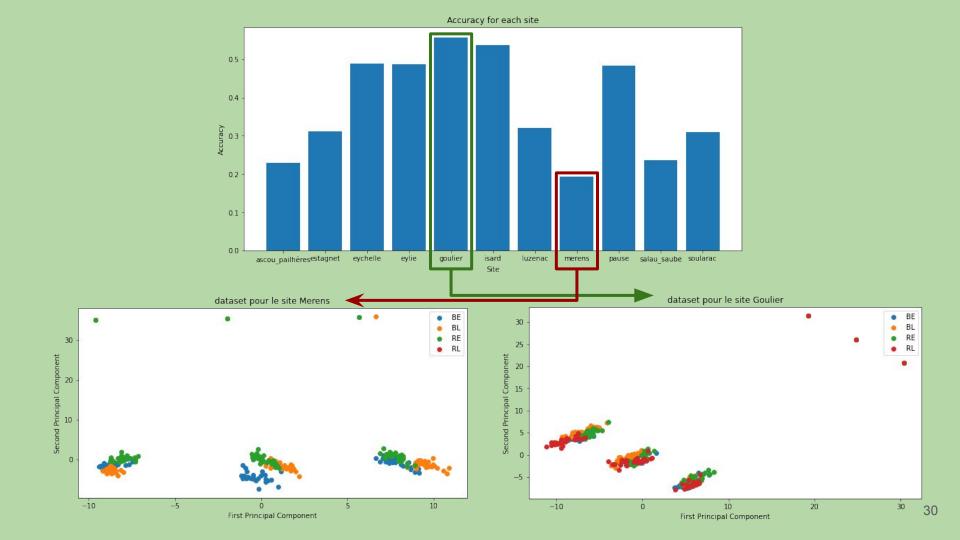






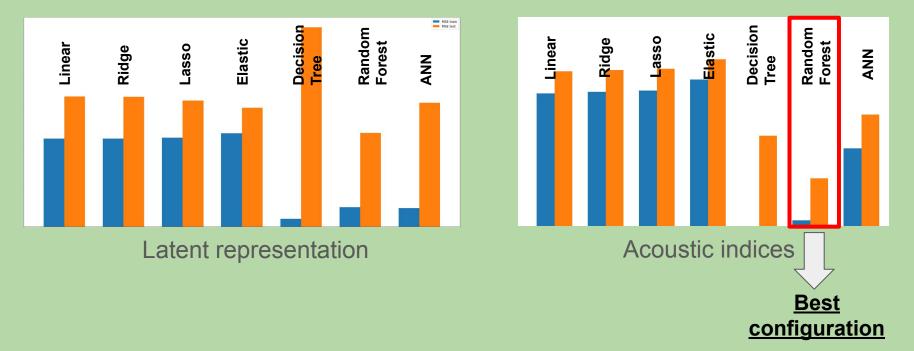
### Analysis of embedded indexes approach



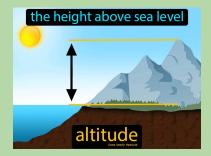


### Environmental Analysis: Choice of model and representation

MEAN MSE Benchmark (Blue: Train; Orange: Test)



### **Environmental Analysis: Best configuration Evaluation**



Average altitude (with an average accuracy of ~50m)



Distance to a trail (with an average accuracy of ~2m)



Slope (with an average precision of ~2°)



dominant species 87% ACC (3 categorical labels)

# Perspectives

What are the relevant research perspective?

- More accurate latent space representation
- Train/Test systems on a wider range of data (geographic locations, seasons, weather, time of day, etc.)
- Cross-referencing results to understand environmental changes (global warming, increasing forests etc.)

# Questions