PROGRESS 3: CNN - BIRDCLEF 2016 BIRD SONG IDENTIFICATION

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Outline

- Recap
- Literature Review
- Methodology
- Preliminary Result
- Next step

Recap Introduction

- trained on 11 bird sound class with manual labeled 120 wav file (10 per class)
- \rightarrow download all sound of 11 class with API \rightarrow 1,847 file (100-200 per class)
- goal: find the model that work on this data before append number of class



Recap

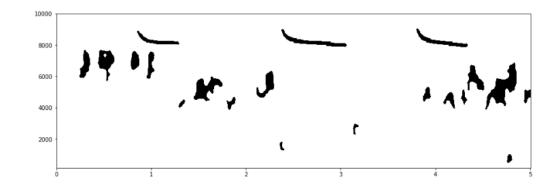
Training data problem

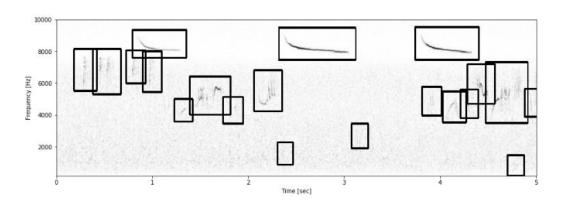
- label only 1 class per file but want multi-label result
- mp3 to wav stereo (multi-channel) sound: node-lame → mean
- quality of sound: using all / only A (~701) / A & B (~1,600)
- sound type (call/song): combine / separate class
- API download problem

Recap

Last algorithm problem

- former model: feature = max correlation with segments in spectrogram
- 4,143 feature \rightarrow 67,991 feature
- prediction time \sim 1 times of sound length \rightarrow \sim 20 times
- too slowly !!!!!



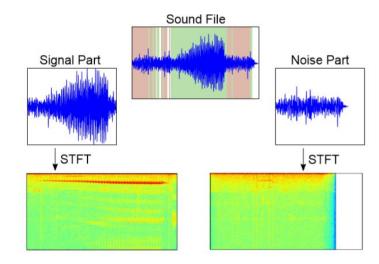


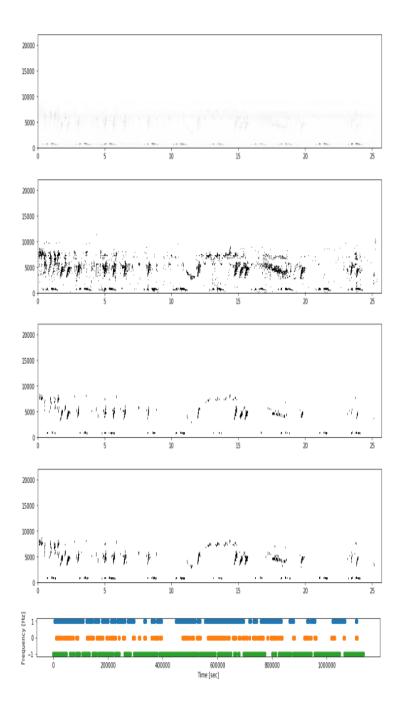
Literature review

- Audio Based Bird Species Identification using Deep Learning Techniques
 (Elias Sprengel and teams, 2016)
- winning solution of <u>BirdCLEF 2016</u>
- data 33,000 recording with 999 difference species (25 files/class)
- mean average precision score of 0.686 (include background species 0.555)

Feature generation

- signal / noise separation
- STFT \rightarrow normalize \rightarrow smooth \rightarrow median clipping
- mask vector for separate audio files





Feature generation

- dividing the spectrograms into chunks
- fixed size input of 512 pixel (~ 3 sec)
- each chunk is unique sample (because separated signal / noise)
- multiple prediction per file \rightarrow average(?) to final

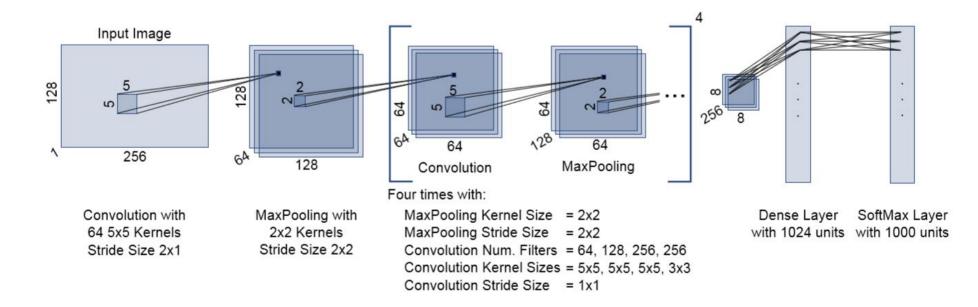
Feature generation

- data augmentation > avoid overfitting
- time shift
- pitch shift
- combining same class
- adding noise

	MAP (FG only)	MAP (FG & BG)
Baseline	0.842	0.728
w/o Noise	0.831	0.731
w/o Same Class	0.839	0.730
w/o Time Shift	0.801	0.701
w/o Pitch Shift	0.828	0.725
w/o Noise and Same Class	0.768	0.661

Network architecture

- 5 CNN and 1 dense with max pooling and ReLU activation function
- batches of 8 / 16 examples
- train: test = 90:10



Preliminary result

- signal part audio file (5 Gb) on Google drive
- trained only A/B quality files
- using Pytorch on Google Colab
- there was some bug...



```
Saving..
1 / 20 Batch Loss: 8,056486402665199
Saving..
2 / 20 Batch Loss: 8.033975657492153
Saving..
3 / 20 Batch Loss: 8,033110983582379
4 / 20 Batch Loss: 8.03337495861993
Saving..
5 / 20 Batch Loss: 8.032938653365814
   20 Batch Loss: 8.068400501552459
    20 Batch Loss: 8.110126499089532
    20 Batch Loss: 8,11012649685261
    20 Batch Loss: 8,110126496106968
   / 20 Batch Loss: 8.11012649685261
11 / 20 Batch Loss: 8,110126495361328
12 / 20 Batch Loss: 8,110126495361328
13 / 20 Batch Loss: 8.110126495361328
    20 Batch Loss: 8.110126495361328
15 / 20 Batch Loss: 8,110126495361328
16 / 20 Batch Loss: 8.110126495361328
    20 Batch Loss: 8.110126495361328
    20 Batch Loss: 8.110126495361328
    ' 20 Batch Loss: 8.110126495361328
20 / 20 Batch Loss: 8.110126495361328
== Training Finish ==
```

Next Step

- find the best model on 11 class: good AUC / recall / prediction time
- write UI to use on mobile phone
- define new scope and test
- test with real data (manual record / data from other website)
- test in the park on mobile phone

Experiment

BirdCLEF 2019



- Task1 Bird species detection in soundscapes
- Task2 Birds counting in soundscapes
- Jan 2019: training data release on <u>www.crowdai.ora</u> → updated on 20 Feb !!
- March 2019: test data release
- 1st of May 2019: deadline for submission of runs by the participants
- 13th of May 2019: release of processed results by the task organizers
- 24th of May 2019: deadline for submission of working notes papers

