

ICML 2013 Whale Challenge Method Documentation

Yun-Chiao Lee and Pei-Hao Su

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Foreword

- We survey and found the key factor of Kaggle winner last year is **Template Matching**!
- Therefore we modified his released code to fit our data
- Surprisingly, the winner of the last year win this year's ICML whale challenge again with same methods and more positive template
- Small conclusion: image processing methods are useful for whale sound detection!

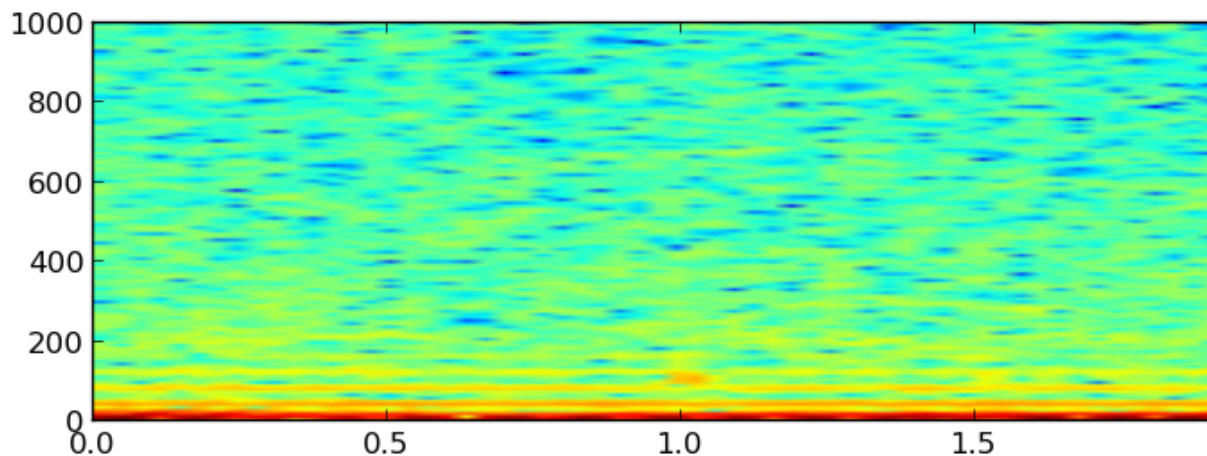
Whale Sound detection

- Feature extraction:
 - Template Matching
- Classification Algorithm:
 - Gradient Boosting Classifier
- Language:
 - python
- Tool:
 - opencv package, scikit learn package

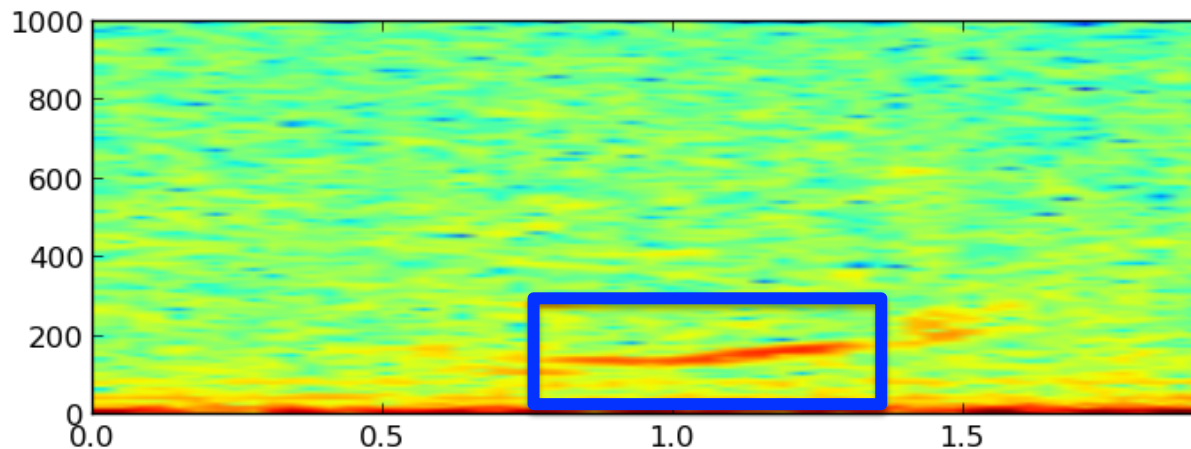
Spectrogram

params = {'NFFT':256, 'Fs':2000, 'noverlap':192}

- Clip without whale sound

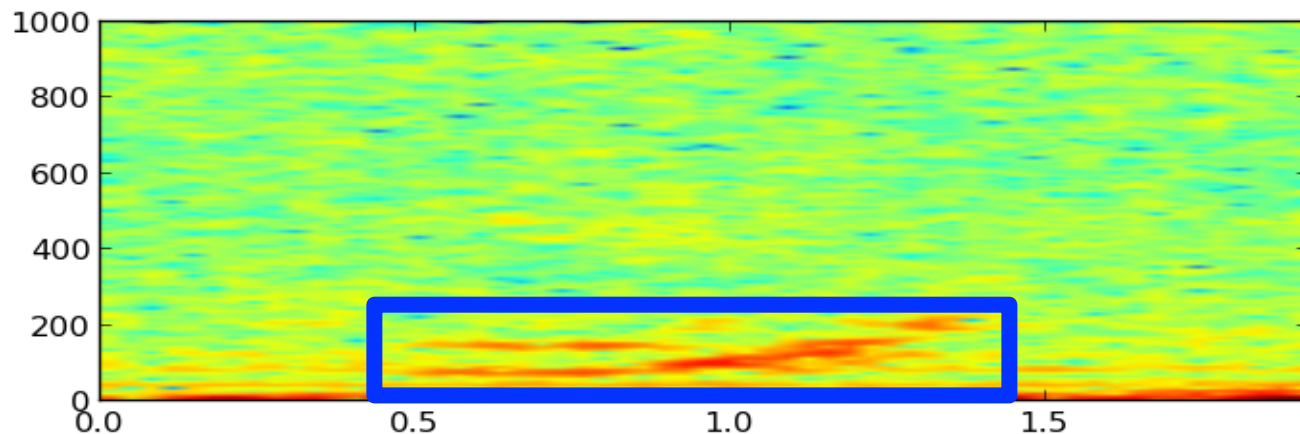


- Clip with whale sound



Preprocessing

- Cut margin
 - We choose clip with whale sound and cut the margin of positive part within spectrogram manually (blue part)
 - Totally 30 manual templates



- Contrast enhancement on each spectrogram to make sound part more explicit

Template Matching

- Up-sweep trajectory in spectrogram
 - Calculate Max correlation and find X,Y location
centTime, bwTime, skewTime, tvTime (centroid,
width, skew, and total variation)...
- We found that most whale sounds spread at frequency between 100-250Hz
- Some audio clips had very low SNR ratio

Classifier

- 4-fold cross validation
- We tried several machine learning classifiers
 - 3-Nearest Neighbors
 - Linear SVM ($C=0.025$)
 - RBF SVM ($\gamma=2, C=1$)
 - Decision Tree ($\text{depth}=5$)
 - Random Forest($\text{max_depth}=5, \text{n_estimators}=10, \text{max_features}=1$)
 - Naive Bayes
 - LDA
 - QDA
 - Gradient Boosting (ensemble of 15000 decision tree)
- Gradient Boosting yields the best result!