

Pipe Organ Stop Identification via Non-Negative Matrix Factorization Autumn Coe | ECE 477 Computer Audition

Abstract

- Most instrument identification methods (Cepstrum Coefficients, Source Filter Models) rely on the uniqueness of harmonic peak amplitudes.
- These methods are vulnerable to multiple instruments playing notes in unison or octaves due to the overlap of harmonic peaks.
- Since multiple stops are frequently turned on together, pipe organs frequently have different stops playing in unison or octaves.
- In this project, sparse NMF is used to take advantage of the additive linearity of the fourier transform.

Objective

- Train a note dictionary with non-negative matrix factorization (NMF).
- Extract dictionary note activations in a sparse matrix using the LASSO algorithm.
- Identify stops by grouping activations by stop and thresholding the maximum activation value of each stop.
- Display results in an intuitive manner using a MatLab GUI

Pipe Organ Overview

- Pipe organs contain multiple stops that imitate orchestral instruments. The main families are: Flutes, Principals, Strings and Reeds.
- Stops can be turned on individually or simultaneously so that multiple unison or octave notes will sound when a single key is pressed.
- Different stops can be selected for each keyboard.

NMF

• NMF is used to decompose a spectrogram (V) into a dictionary of notes (W) and and activations (H) such that:

$$W \times H \approx V$$

• W and H are updated using euclidean distance multiplicative updates [1] such that:

$$W_{ia} \leftarrow W_{ia} \frac{(VH^T)_{ia}}{(WHH^T)_{ia}}$$

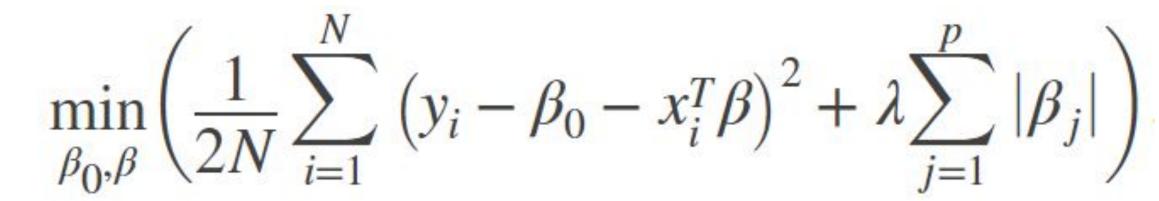
$$H_{a\mu} \leftarrow H_{a\mu} \frac{(W^T V)a\mu}{(W^T W H)_{a\mu}}$$

NMF Continued

- A note dictionary (*W*) is created for each stop with *a* equal to the number of notes in that stop. Dictionaries are trained on chromatic scales played on each stop individually.
- Individual stop dictionaries are concatenated to create a complete note dictionary.
- Just using NMF for stop identification produces an activation matrix that is too noisy as shown in fig. 1. To prevent this, a sparsity algorithm is employed.

LASSO Algorithm

• To calculate the sparse activation matrix, the LASSO (least absolute shrinkage and selection operator) algorithm provided in MatLab [2] is used:



- *N* is the number of observations
- \circ y_i is the response at observation *i*
- $\circ x_i$ is a data vector of length p at observation i
- \circ λ is a nonnegative regularization parameter
- $\circ \beta$ is a vector of length p
- The LASSO algorithm was applied to the spectrogram as follows: \circ x = W, the trained note dictionary
- $\circ y = V_{y}$, each frame of the spectrogram $\circ \beta = H_{\mu}$, each column of the activation matrix
- This resulted in a more accurate activation matrix as shown in fig.

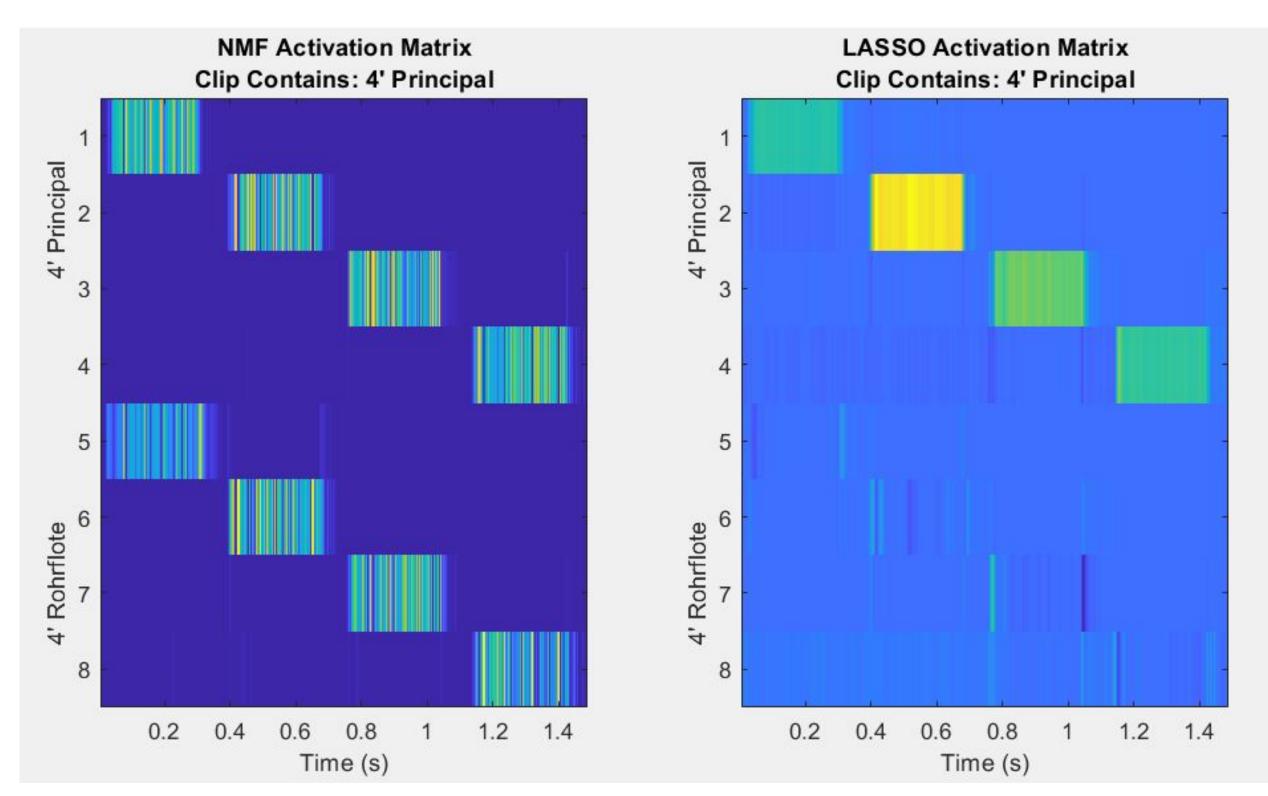


Figure 1 NMF vs LASSO activation matrix.

Stop Identification

- representing each stop.
- fig. 2.

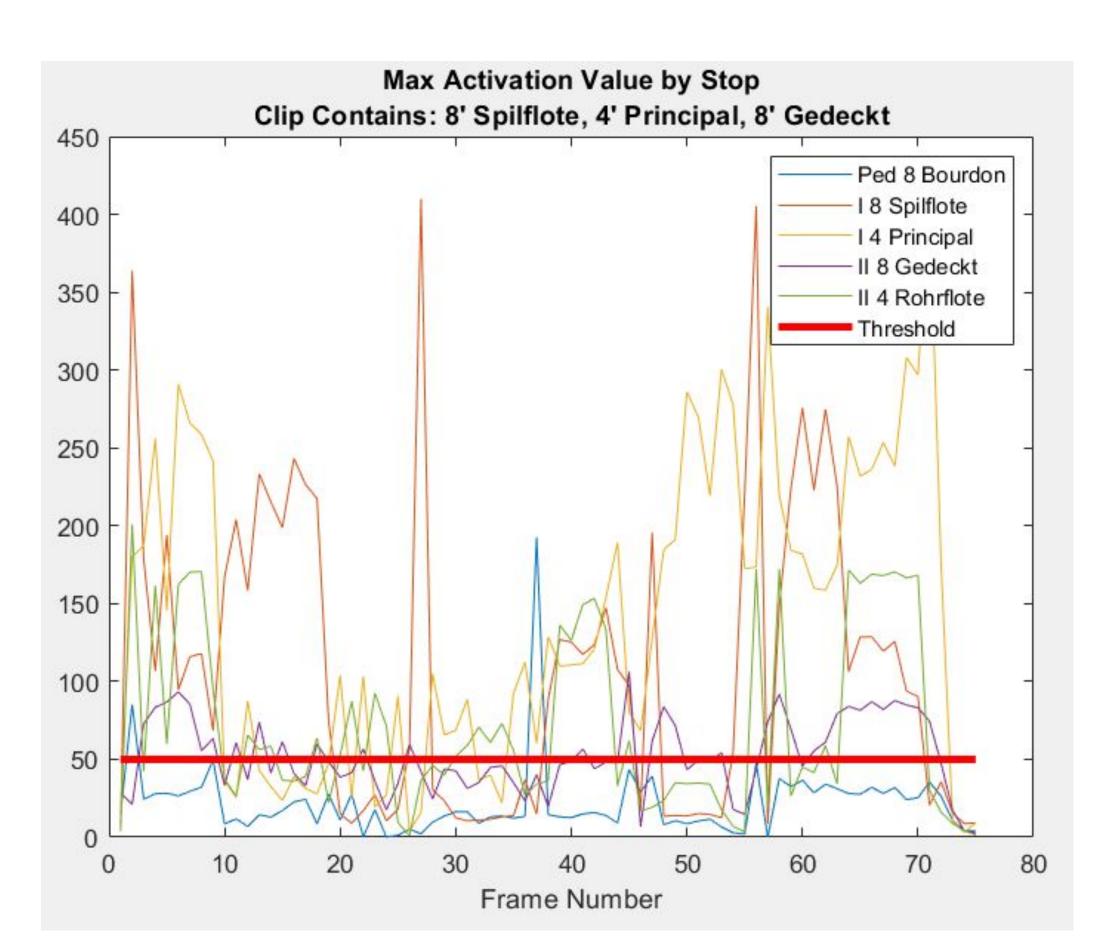


Figure 2: Maximum activations for each stop.

Results and Future Work

- stops on this organ.
- time.

Citations

- Dec. 2018].

• The rows of the activation matrix are divided into sections

• To determine if a stop is being played, the maximum activation value is taken from each stop section and thresholded as shown in

• As shown in fig. 2, the accuracy is poor. This can be attributed to the difficulty of separating unison notes and the similarity of the

• Even with the pretrained note dictionary, the LASSO algorithm is quite slow preventing the stop identification from running in real

• Future work includes improving the accuracy by refining the note dictionary and expanding the project to larger pipe organs.

[1] D. Lee and H. S. Seung, "Algorithms for non-negative matrix factorization," in Proceedings of the Neural Information Processing Systems (NIPS), Denver, CO, USA, 2000, pp. 556–562.

[2] Mathworks.com. (2018). Lasso or elastic net regularization for linear models - MATLAB lasso. [online] Available at: https://www.mathworks.com/help/stats/lasso.html [Accessed 4

