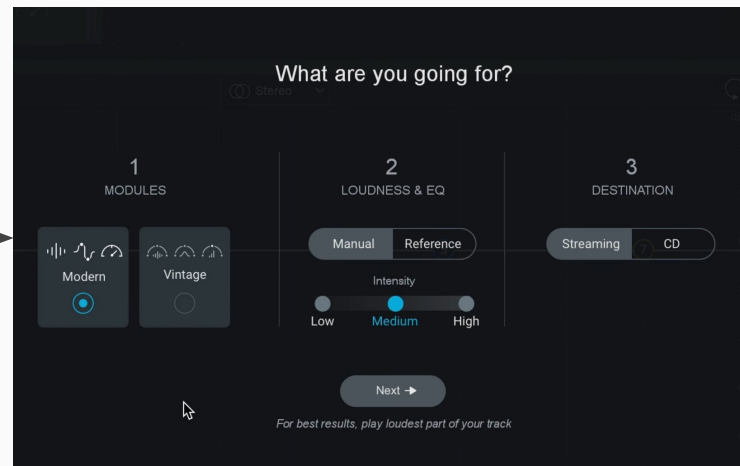
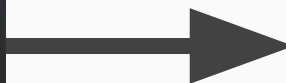


# Ratio Estimation for Dynamic Range Compressors

José Díaz Rohena & Wesley Yu



# Automatic Control Through The Ages



# Related Work

1. Parameter Automation in a Dynamic Range Compressor (2013)
  - a. High level user control—adapts to input signal
  - b. Real time, time domain features
2. Feature Selection for Dynamic Range Compressor Parameter Estimation (2018)
  - a. Using reference audio to set parameters
  - b. Variety of time domain and frequency domain features

# Dataset

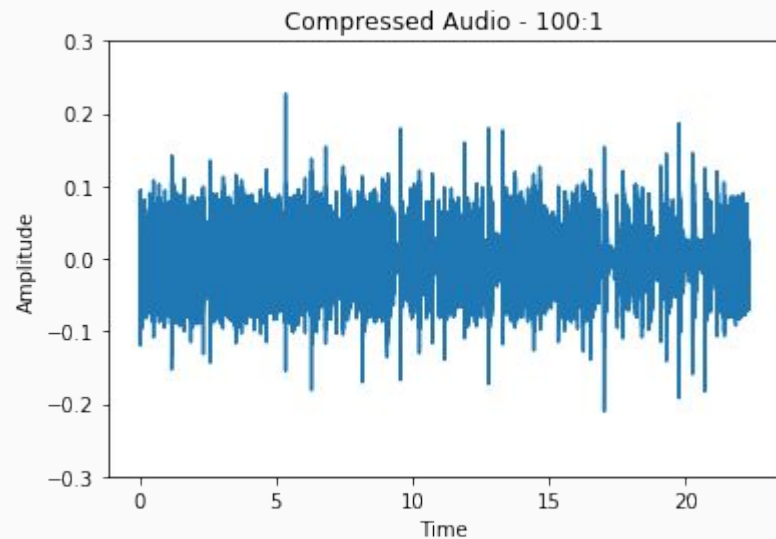
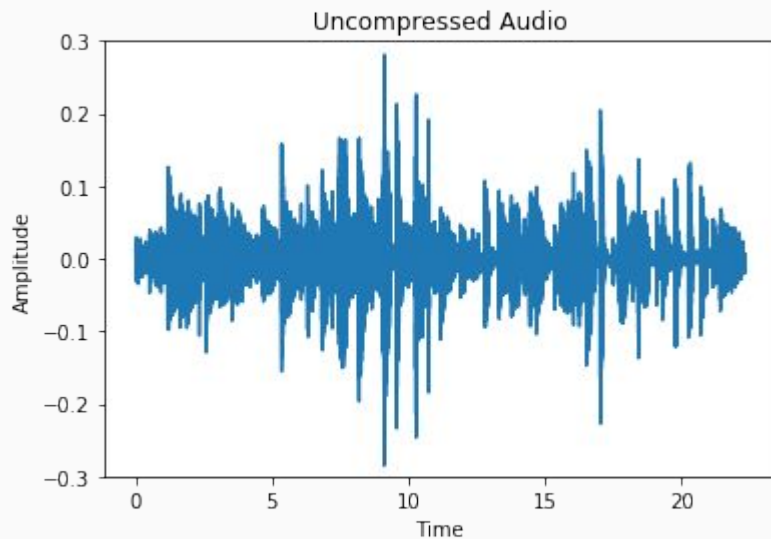
## 1. GuitarSet

- a. “Mic” subset
- b. 360 examples
  - i. Split into 5 second segments
- c. Good mix of transient and tonal information

## 2. Compression

- a. Segments compressed at random ratios, then features are extracted
- b. Feedforward, Peak detector
- c. Ratios = 1:1, 8:1, 100:1, Threshold = -40 dB, attack = 1.3 ms, release = 300 ms, knee = 0 dB

# Compression Visualized



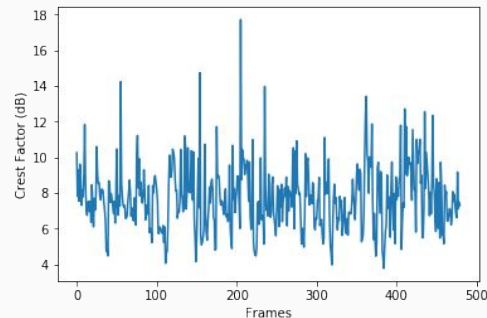
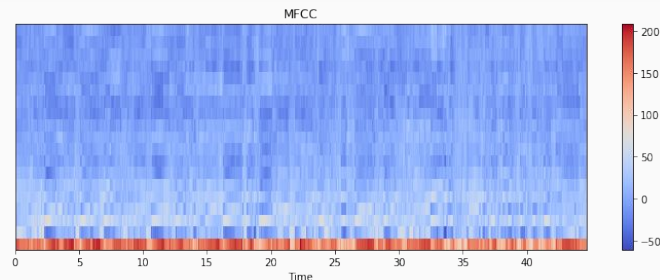
# Features

## 1. MFCC

- a. Timbre

## 2. Crest Factor

- a. Dynamic Range—ratio of peak level to RMS level
- b. Frame Length = 2048 samples, no overlap
- c. Converted to dB



# Training and Evaluation

## 1. 2 datasets

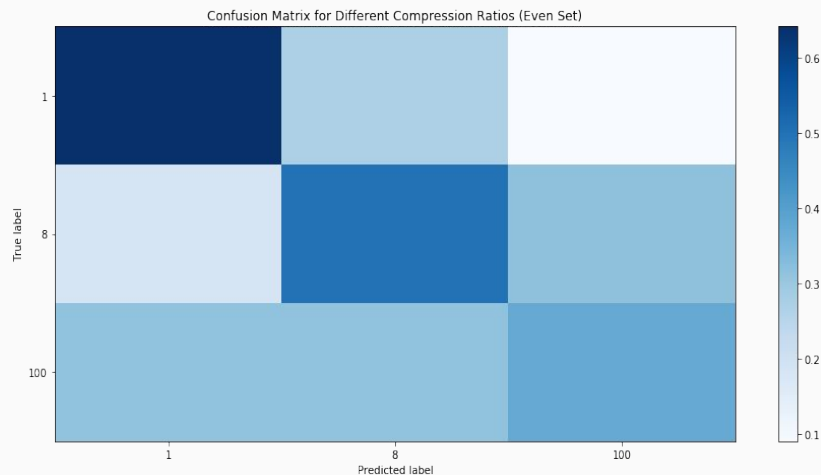
- a. “Even” - 668 1:1, 666 8:1, 666 10:1
  - i. training: 1600, validation: 200, training: 200
- b. “Uneven” - 2004 1:1, 666 8:1: 666 10:1
  - i. training: 2670, validation: 333, test: 333
  - ii. would overlap / more data help?

## 2. scikit-learn Nearest Neighbors

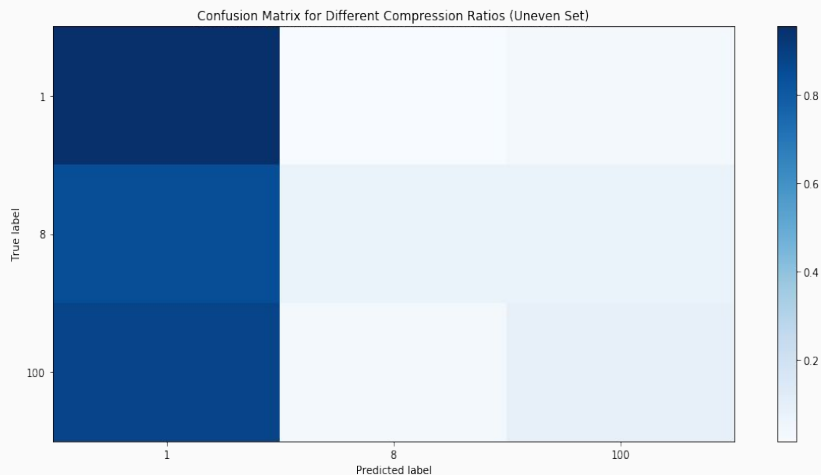


# Results

Accuracy Score = .505  
Balanced Accuracy Score = 0.505  
f1 = 0.50  
k = 50



Accuracy Score = 0.607  
Balanced Accuracy Score = 0.373  
f1 = 0.508  
k = 10





# Discussion

1. How did we do?
  - a. Compressed / not compressed detector
    - i. We need better discrimination
2. Data
  - a. More examples - use all capture methods?
  - b. Use multiple variations of same example
  - c. Split example at note onsets?

Thank You!

