Drum Samples Classification

Final project for MAS 622J/1.126J

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- Objective : Classify individual drum sample
- **Project focus :** Find best method and features set



Classes

- Bassdrum
- Snares
- Hi-Hat
- Cymbal
- Clap
- Tom

Methods

- Support Vector Machine
- K Nearest neighbors
- Neural Network

Features

- **Pitch** Average maximum FFT bin frequency
- **Decay** Time from peak to 50% amplitude
- Brightness high frequency energy
- Rolloff low frequency energy
- **Roughness** based on the frequency ratio of each pair of sinusoids
- Irregularity degree of variation of successive spectrum peaks
- MFCC Mel frequency cepstrum coefficients

Training and testing dataset distributions

| | Tra | ining | Testing | | | | |
|----------|---------|------------|---------|------------|--|--|--|
| | Samples | Percentage | Samples | Percentage | | | |
| Bassdrum | 337 | 33.2% | 224 | 18.9% | | | |
| Snare | 259 | 25.5% | 251 | 21.2% | | | |
| Hi-hat | 195 | 19.2% | 403 | 34.0% | | | |
| Cymbal | 116 | 11.4% | 68 | 5.7% | | | |
| Clap | 97 | 9.5% | 52 | 4.4% | | | |
| Tom | 12 | 1.2% | 187 | 15.8% | | | |
| Overall | 1016 | 100% | 1185 | 100% | | | |

Matlab GUI

- Training and testing sets management
- Saving, loading, feature extraction
- Model management
- Feature space view
- Progress bars

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| -5 -5 | -4 | -3 | -2 | -1 | 0 | 1 | 2 | 3 |
| | X Axis | | | Y Axis | | | | |
| - | brightness | | | fcc2 | \$ | | Plot | |
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K Nearest Neighbors

- Tested 1 < k < 15 with k taking odd values
- Leave-one-out validation
- Forward feature selection

Results : k = 9 using

- Brightness
- Irregularity
- Decay
- MFCC 1
- MFCC 2
- MFCC 3
- MFCC 5

Neural Net

- Used Matlab's neural net tools, including validation
- Performance estimation through MSE
- Tested 1 and 2 hidden layers
- 5:12 units per layer

Results

2 hidden layers with 9 and 7 units respectively

Selected features :

- Rolloff
- Brightness

• Mfcc 1

• Pitch

- Roughness
- Irregularity
- Decay

- Mfcc 3
- Mfcc 4
- Mfcc 7

Support Vector Machine

- Used Matlab built-in SVM tools
- Leave-one-out validation
- One SVM per class One-Versus-All
- Forward feature selection

Results

| | Brightness | Rolloff | Irregularity | Roughness | Decay | Pitch | Mfccl | Mfcc2 | Mfcc5 | Mfcc7 | Mfcc I I | Mfcc13 |
|----------|------------|---------|--------------|-----------|-------|-------|-------|-------|-------|-------|----------|--------|
| Bassdrum | ~ | | | | | | ~ | | | | | |
| Snare | ~ | ✓ | | | | ~ | ~ | | ~ | | ~ | |
| Hi-hat | | | | | ~ | ~ | | | | | ~ | |
| Cymbal | ~ | | | ✓ | ~ | | | | | ~ | | ~ |
| Clap | | | v | | | | | ~ | | | | |
| Tom | | ~ | | | | | | | | | | |

Performance



Improvements to be made

- Extend and manually validate training set
- Try temporal features approaches, namely HMM
- Implement multiple algorithms and majority selection

Future Work

- Subjective features extraction and sample retrieval
- Melodic instruments samples