Music Clip Identification using Randomized LSH Tables

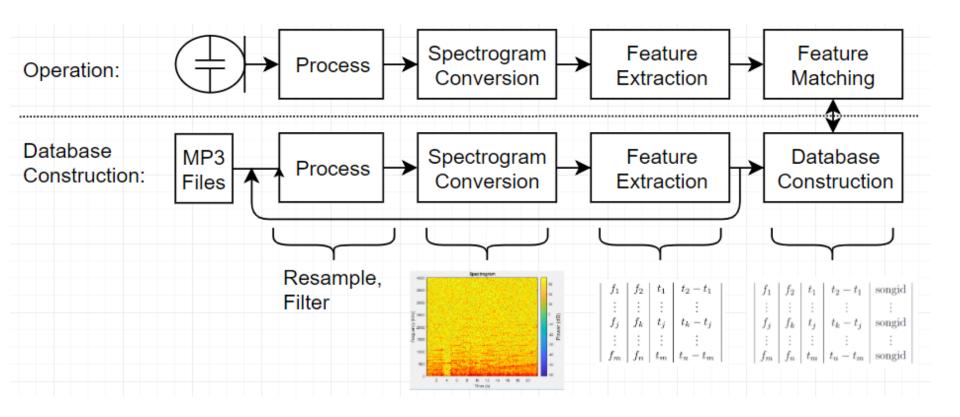
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Problem Statement and Motivation

- Find the identity/source of a song given a small clip (<30 seconds)
- Interesting and nontrivial due to large amounts of detail in an sound wave, as well as random initial conditions and exogenous factors
- Development of these algorithms is motivated by people interested in discovering new artists/music

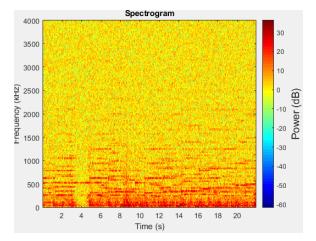
General Operation

 There seem to be many opportunities to implement RAs here



Spectrogram and Feature Selection

- Frequency and amplitude variations wrt time. Eg: Fallout 4 theme by Inon Zur
- Features first filtered by amplitude on a log scale, filtered to 30 per second, sort by magnitude, group by fanout distance.
- Frequency pairs are generated, with their coupled time differences into a single table per song





Shingle Generation and minHash function

 First a "shingle" is generated using a conventional hashing mechanism.

$$x = T(:,2) + 2^8 T(:,1) + 2^{16} T(:,4)$$

$$h(x) = \min(\boldsymbol{a}x + \boldsymbol{b} \mod w) \in \mathbb{R}^k$$

- Then, we perform minHash on this shingle, for w being largest next prime number.
- Dimension of x can be anything, hash output dimension is fixed
- The coefficients a and b are randomly chosen integers less than the max value of shingle x

Matching Algorithm

- Just make a table and corresponding hash sequence of the clip, then search the database, taking advantage of locality sensitivity
- Find the highest value per table, and return ID

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Algorithm 1 Simple Matching Function
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1: Cliphashtable = minhash(make\_table(clip));
2: k = const, dim(h(x))
3: clip\_score = 0
4: for i = 1:N do
5: hashTable\_sentence = minhashTable((i-1)*k+1:i*k,1)
6: local\_sens\_bool = abs(Cliphashtable - hashTable\_sentence) \le 100
7: local\_score = \sum (local\_sens\_bool)
8: if song\_score < local\_score then
9: song\_score = local\_score
10: songName = i
11: end if
12: end for
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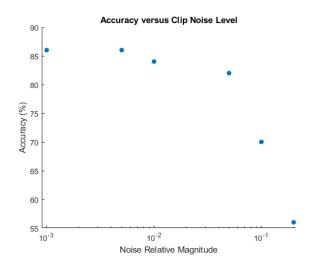
Alternative Approaches and Obstacles

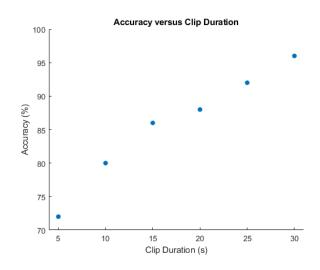
- Various LSH Algorithms
 - MinHash
 - Euclidean Norm Distance Hash
 - SimHash
- Various Matching Algorithms
 - K nearest neighbours
 - Direct Equality (ends up being same as epsilon ball)

Plots and Demonstration



- Savings in required table size
 - Fifty test songs ~2 GB
 - Key Features ~200 kB (0.01%)
 - Randomized minHash table ~50 kB (0.0025%)





Final Results and Conclusions

For k=500, low noise = \sim 0.1*A, high noise \sim 1*A

- Detect 66% of 5 second clips with low noise
- Detects 58% of 5 second clips with high noise (basically static)
- Detects 88% of 15 second clips with low noise
- Detects 94% of 22 second clips with low noise
- MinHash algorithm works well, having static database entries for varying spectrogram thresholds is very useful

Future Work and Extensibility

- Alternative methods for association and classification
- Pulsar detection from radio observations
- Performing the lost-in-space attitude problem with an image of stars and a star catalog
- Many more

Questions?



References

- Original Shazam paper
- Hashing for Similarity Search: A Survey
- Near Optimal Hashing Algorithms for Approximate Nearest Neighbor in High Dimension
- Near Duplicate Image Detection:min-Hash and tf-idf Weighting
- Computer Vision for Music Identification