

A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light green. They are positioned diagonally, with the blue one partially covering the green one.

Music Genre Classification

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Our Goal

- To build a predictive model to classify the genre of a given song into 1 of 10 genres
 - Genre - the type or style of music
- Take a 30 second audio clip and classify its genre using our model



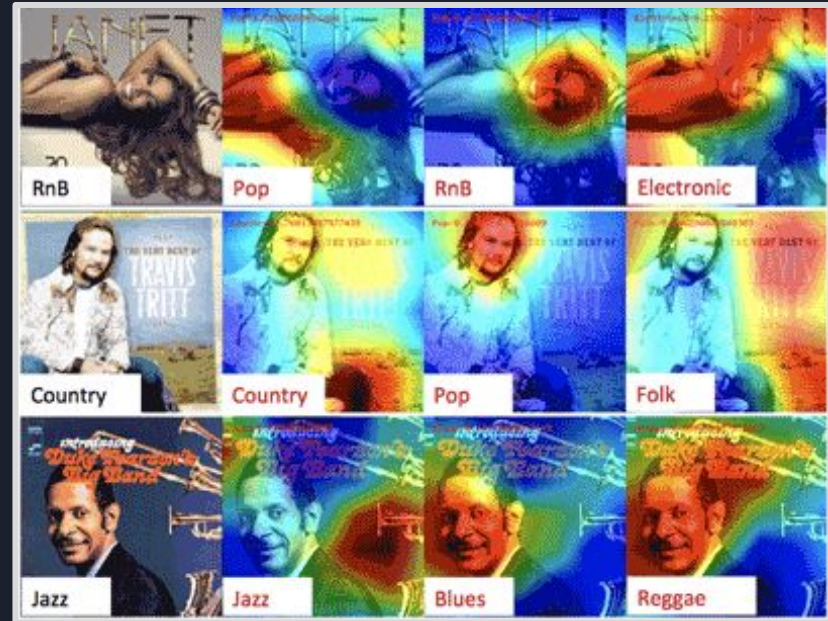
Importance of Music Genre Classification

- Helps music streaming platforms to...
 - Provide relevant recommendations
 - Curate playlists
 - Play relevant songs once our queue is over
 - More accurately label newer & or niche genres



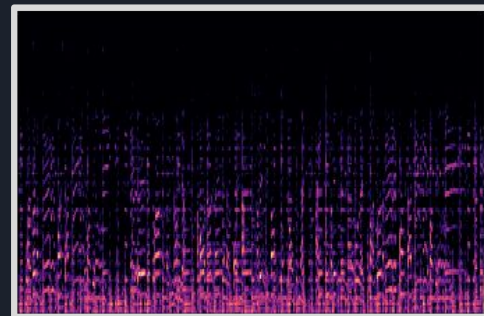
Similar Works

- Multimodal Deep Learning for Music Genre Classification
 - CNN
 - Multimodal approach
 - Single-label
 - Multi-label
 - Album cover



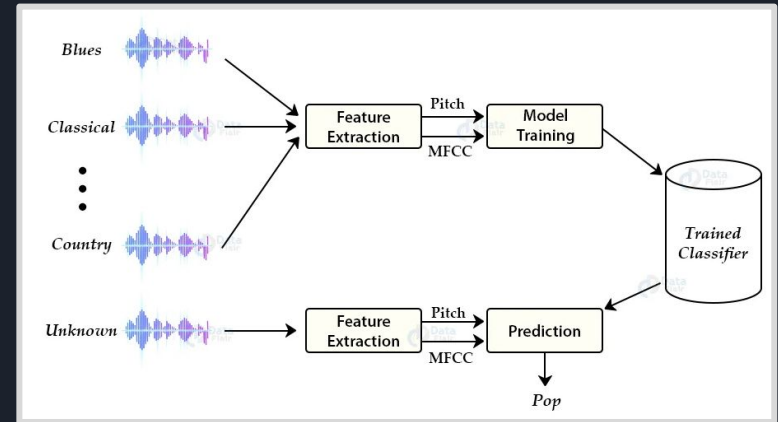
Data

- GTZAN Music Genre Dataset
- 1,000 song samples, each 30 seconds long, belonging to 10 different genres
- Contains
 - 3 second clip features
 - 30 second clip features
 - 30 second audio clips
 - Visual representations of the clips as Mel Spectrographs
- Challenges
 - Extracting proper/sufficient features
 - Restricted number of samples



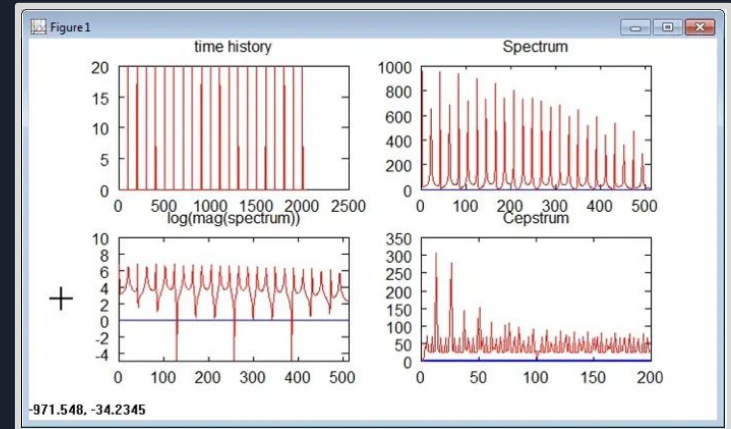
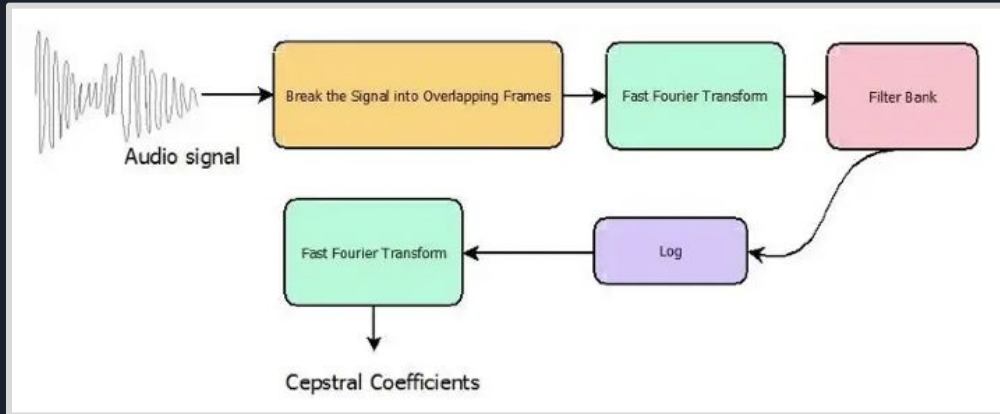
Approach

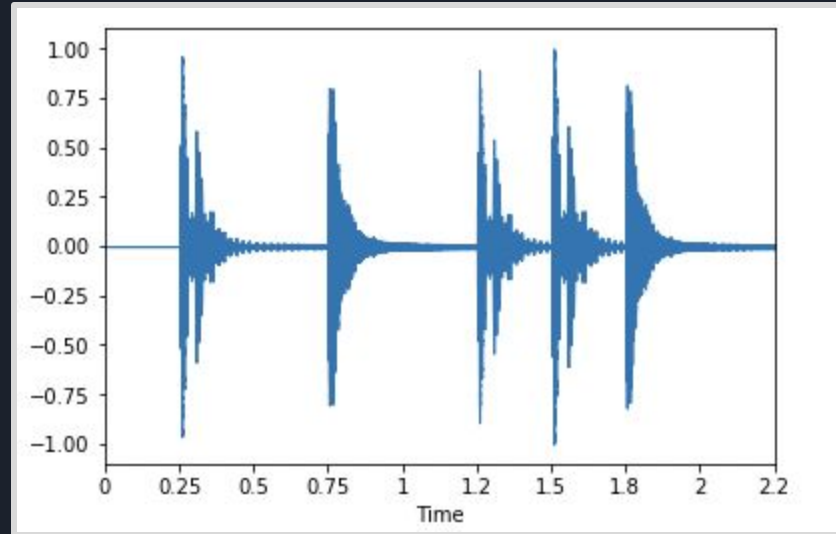
- Decided to go with K-nearest neighbors using MFCC's
- Feature extraction
- Mel Frequency Cepstral Coefficients
 - Divide signals into clips 20-40 ms long
 - Identify frequencies present in each clip
 - Separate linguistic frequencies from the noise
 - Use DCT to separate the noise
- Make predictions using KNN
- Try and match or beat the accuracy of past work



Mel Frequency Cepstral Coefficients (MFCC)

- Cepstrum - rate of change of spectral bands
- Fourier Spectrum
- Quefrequency domain -> cepstrum
- Mel scale - relates perceived frequency of tone to actual frequency
- Any sound generated by humans -> shape of vocal tract





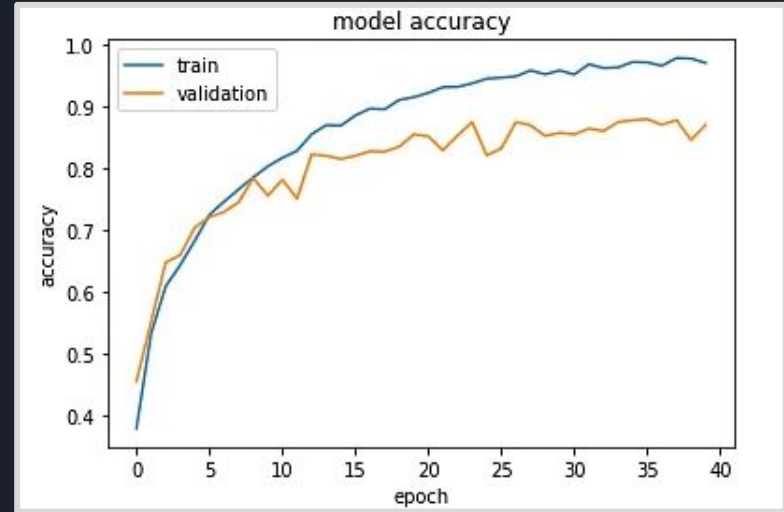


Process/Model

- Get distance between feature vectors and find neighbors
- Identify nearest neighbor
- Evaluate model
- Extract features from data (MFCC's)
- Train the model
- Make predictions using KNN
- Test with new untrained audio file

Results and Performance

- After many tweaks and iterations -> ~70% accuracy on test data
- Pretty good considering size of data and others' scores
- After researching - CNN seems to be more accurate using Mel Spectrographs
 - ~88% accuracy



CNN model accuracy



Conclusion

- In the future - more data could create a better model
- Try using CNN for better results
- Classify songs into more than just 10 genres - more useful/realistic