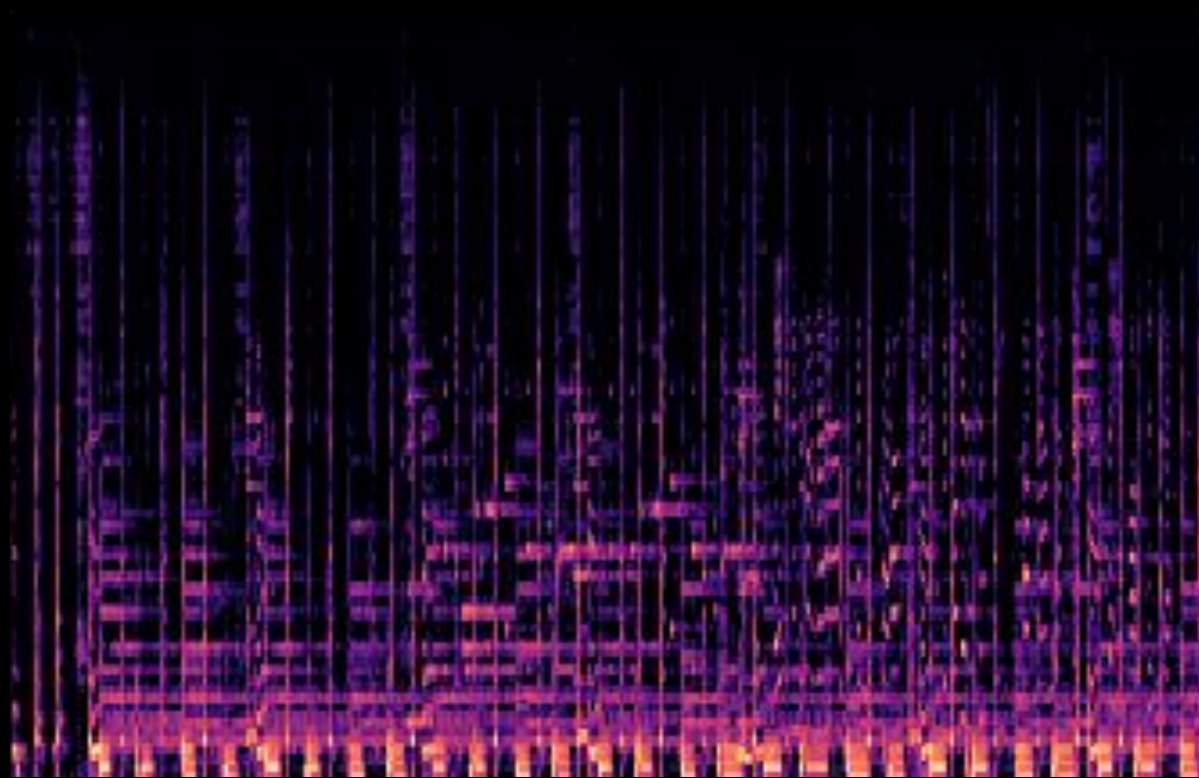




**What Music Genre is shown below?**





# Music Genre Classification from extracted audio data

Aidan O'Keefe





# AGENDA



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Understanding**

**02**

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**06**

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# Business Understanding

A pink audio waveform is positioned in the top right corner of the slide, extending horizontally and slightly upwards.

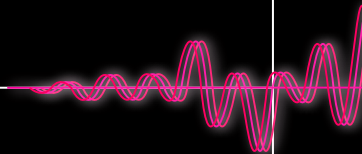
**A DJ has a disorganized collection of audio tracks from recording samples over the years.**

**The DJ wants to be able to sort each of these samples into the appropriate genres to make them easy to catalog and search through.**

**Since the DJ has no information on the samples but their audio files, we have been enlisted to create a classification model that can sort the audio tracks into genres.**

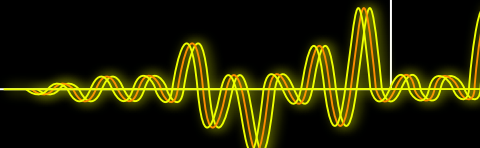
A yellow audio waveform is positioned at the bottom left of the slide, extending horizontally and slightly downwards.

# Data Understanding

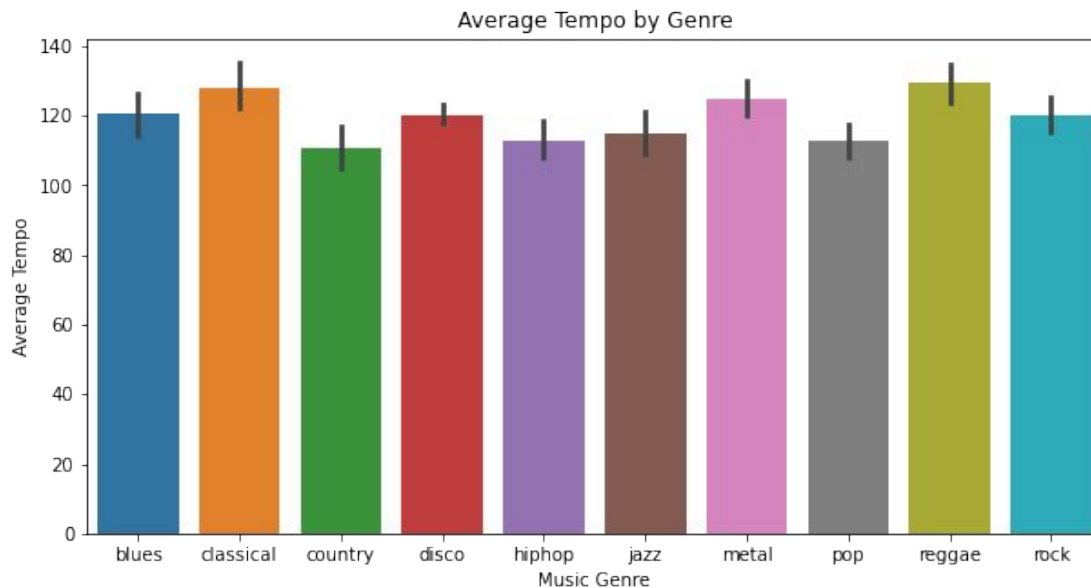


## "GTZAN Dataset - Music Genre Classification" from Kaggle

- 1000 .wav audio tracks of a 30-second duration evenly divided into 10 genres.
- Data extracted using the Librosa python package already available.
- Each row represents an audio track with 60 columns mostly containing a mean and variance computed over multiple features.

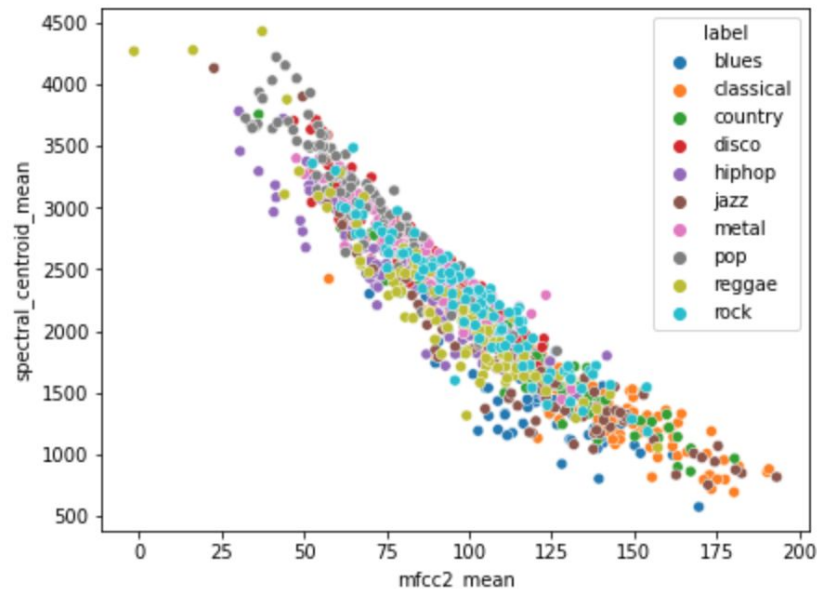
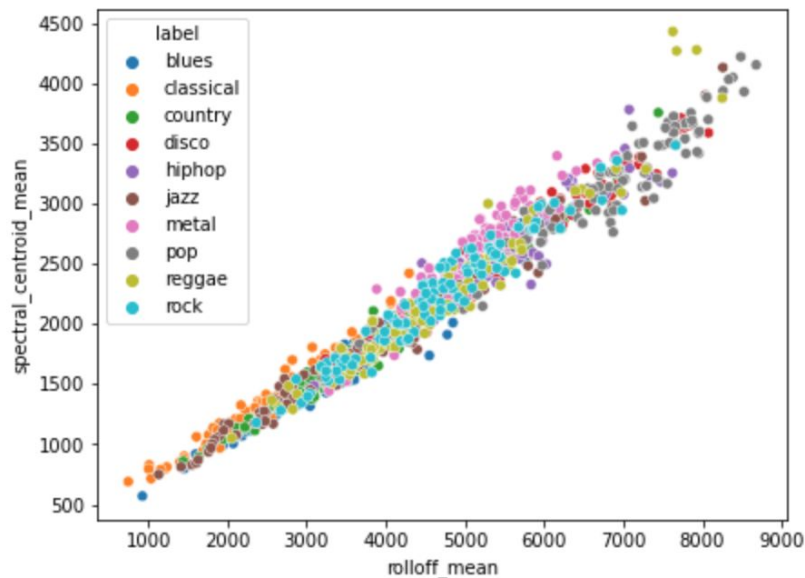


# Data Understanding



# Data Understanding

Predictive Features with some of the strongest correlations.



# Methods



## Data Preparation

- Removed predictive variables with high collinearity
- Scaled data to deal with skewed variables

## Modeling

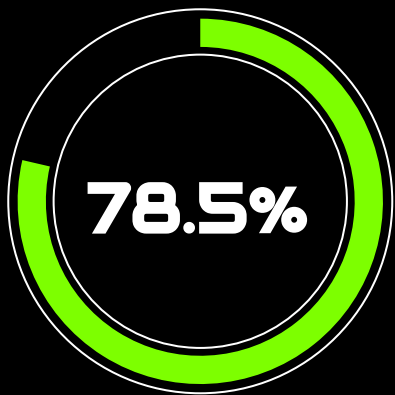
- Implemented and tuned a combination of machine learning models





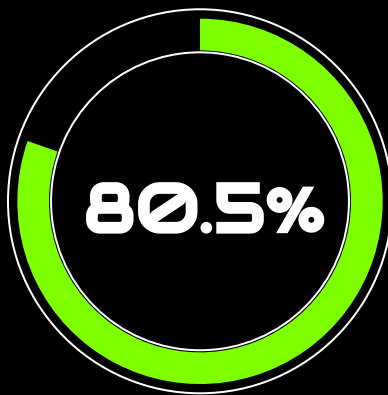
# Results

## Accuracy Scores on unseen data



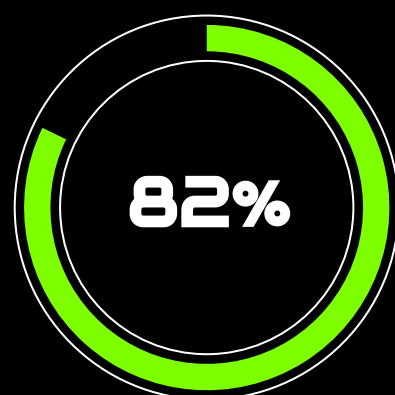
### 3rd Best Model

Untuned Stacked  
LR, XGB, QDA,  
Untuned LR



### 2nd Best Model

Tuned Stacked LR,  
RF, KNN, Untuned  
LR



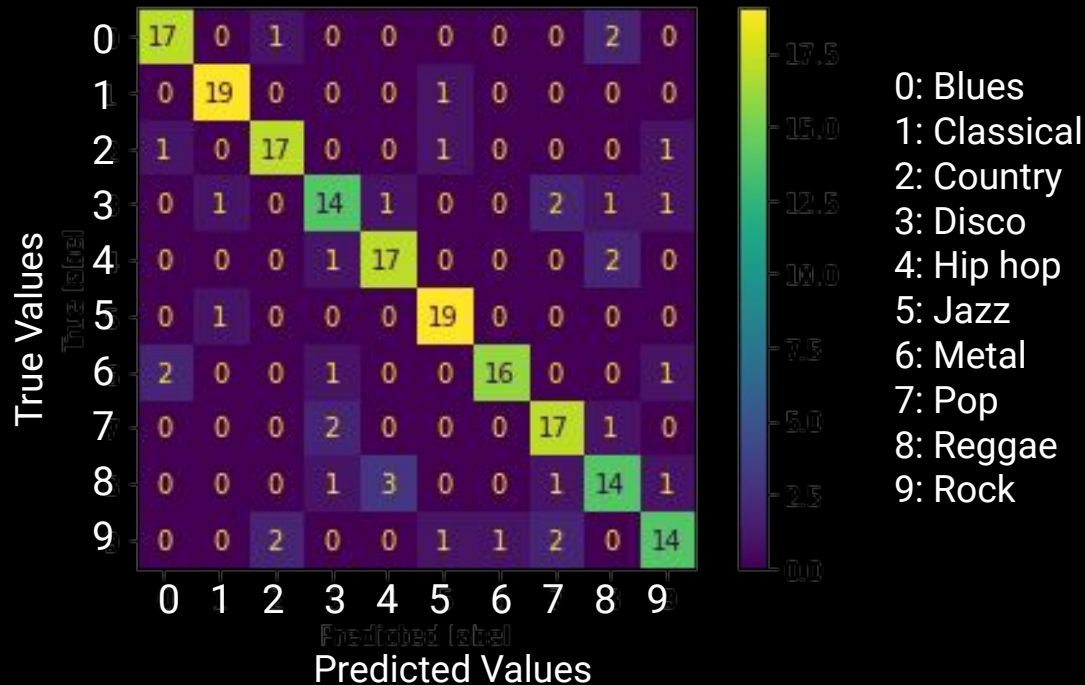
### Best Model

Tuned Stacked LR,  
RF, KNN, XGB, QDA,  
Untuned LR

# Results: Final Model

**82% Test Accuracy**

Tuned Stacked LR,  
RF, KNN, XBG, QDA,  
Untuned LR





# Recommendations




## Work on trouble Genres

Struggled with three genres (Disco, Reggae, and Rock), classifying them correctly only 70% of the time compared to 80%+ accuracy for the other seven genres.



## Get More Data

Using more data could help prevent recurrent overfitting issues and lead to better model performance.



# Next Steps



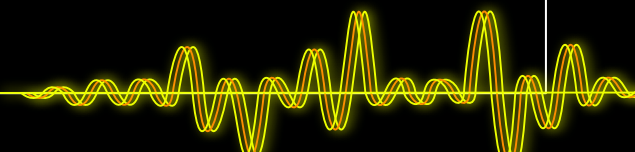
## Add Genres

Adding additional genres would be beneficial to the usefulness and specificity of this model on unseen audio files.



## Extract Feature Importances

Try to better select which columns/features to extract from the audio files in order to reduce complexity and collinearity.



# Thank you

## Questions?

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