

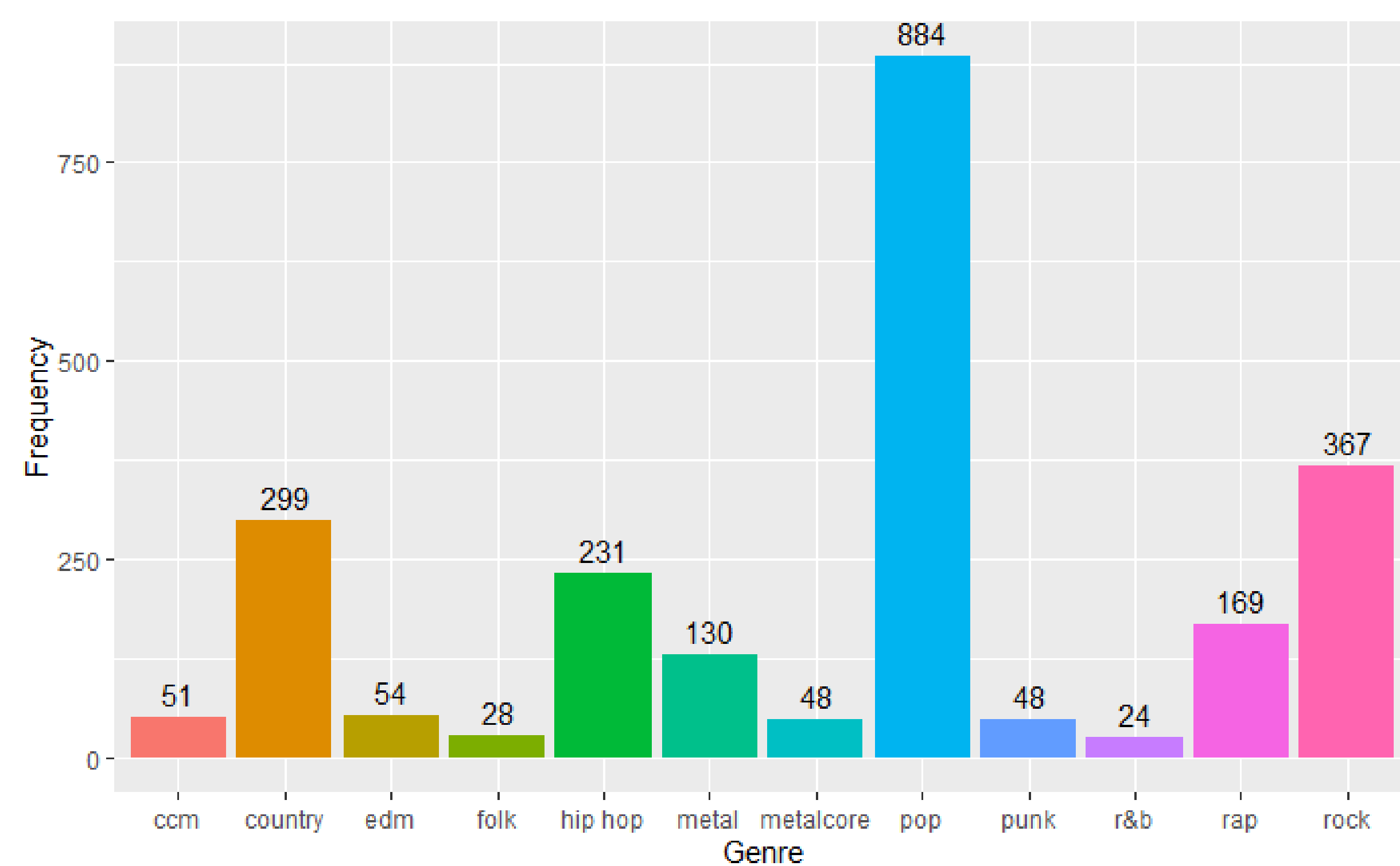
## Abstract

How we define a genre is always changing. Genre is constructed by patterns that we hear in music, but these patterns can change with time, with a movement, and sometimes instantly. The way that we can classify a song by genre is not always based on audio attributes.

The aim of this project is to investigate a potential model for determining the genre of a song track based on numerous audio attribute traits such as song duration, key, mode, time signature, acousticness, liveness, danceability, energy, instrumentalness, loudness, speechiness, valence, and tempo. The models that were constructed include classification trees and support vector machines.

This project was motivated by a simple interest in trying to understand and build connections genre with the song attributes provided by Spotify.

## The Data



**Figure 1: Frequency of Genre in Dataset**

The data was obtained by scraping songlist.com for songs released in 2018 and 2019. The Spotify API was used to collect genre, popularity of the song, and song attributes. Song attributes include song duration, key, mode, time signature, acousticness, liveness, danceability, energy, instrumentalness, loudness, speechiness, valence, and tempo.

After collection, the data was cleaned and all incomplete entries were removed. This provided 3333 data points for analysis. The frequency of genre can be seen in Figure 1.

Note that the frequency of pop songs is much greater than the frequency of any other genre.

## Conclusion

Overall, the support vector machine and classification tree are not perfect and do not produce the best results. In fact, both trees produce an accuracy less than 60% (SVM: 43%, Classification Tree: 46%). In addition, both of the proposed models for determining genre have trouble identifying certain genres (ccm, edm, folk, etc.) and the classification tree, leaves these genres out all together.

## Future Direction

In the future, I would want to obtain a more representative dataset of the music that is released. In addition, I would like to refine the models that I have created for this project and look into other models that would produce better results.

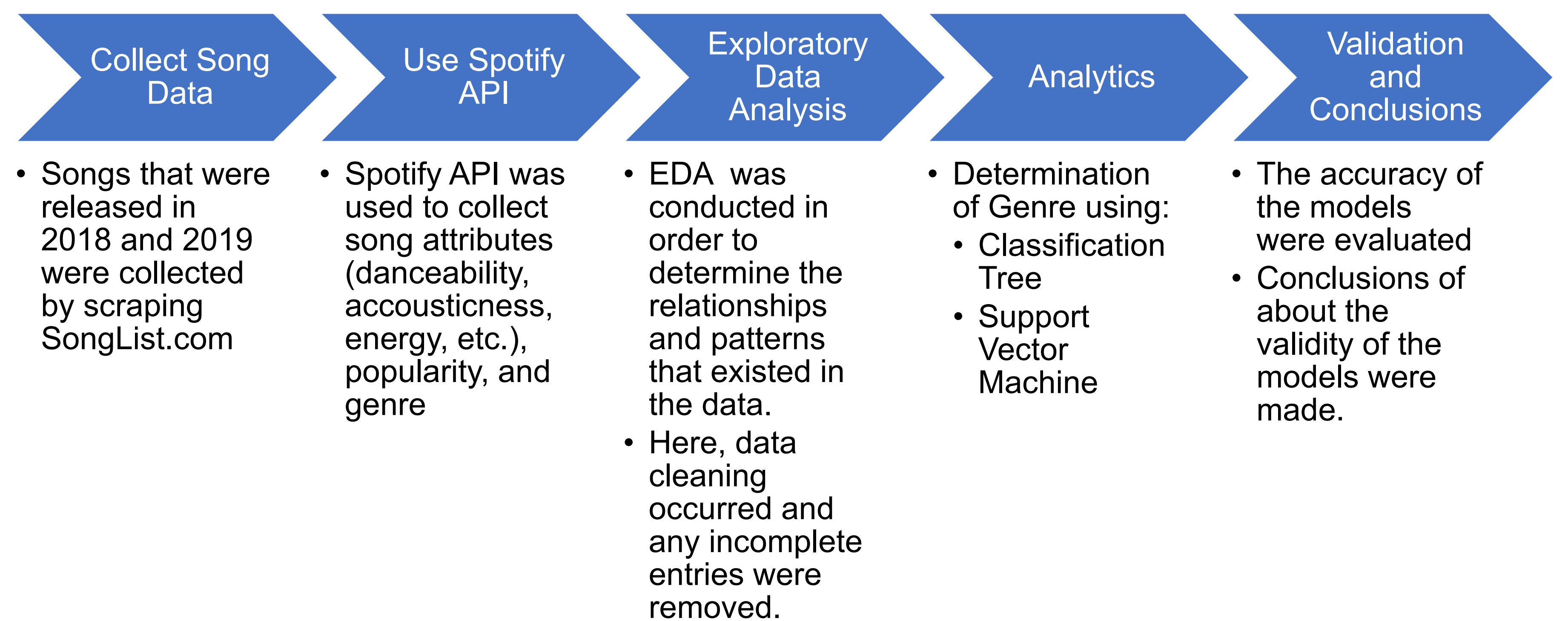
### Resources:

Songlist.com – provided a list of songs released in 2018 and 2019  
Spotify API – used to obtain song attributes

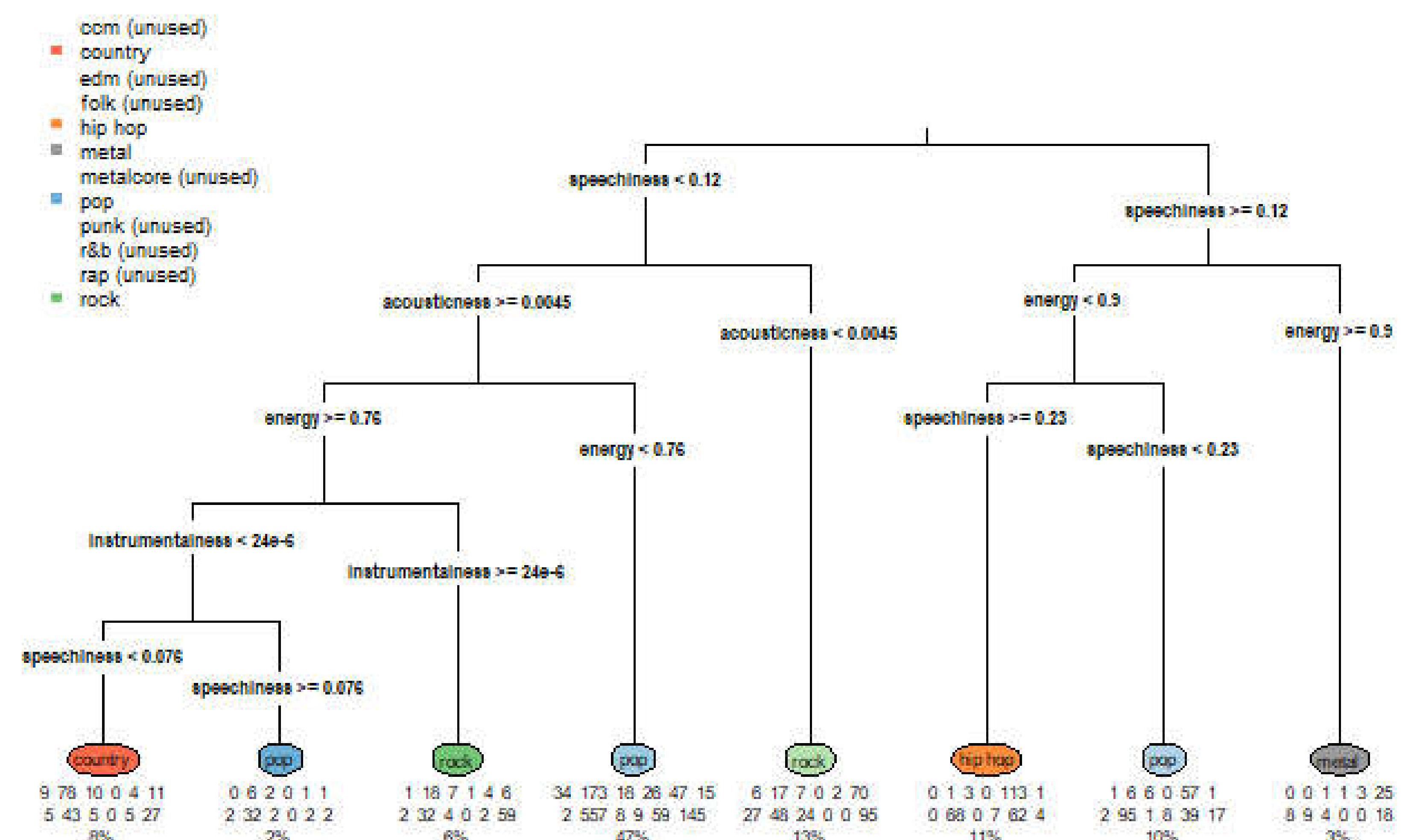
### Glossary:

**API** – Application programming interface  
**SVM** – Support Vector Machine

## Workflow



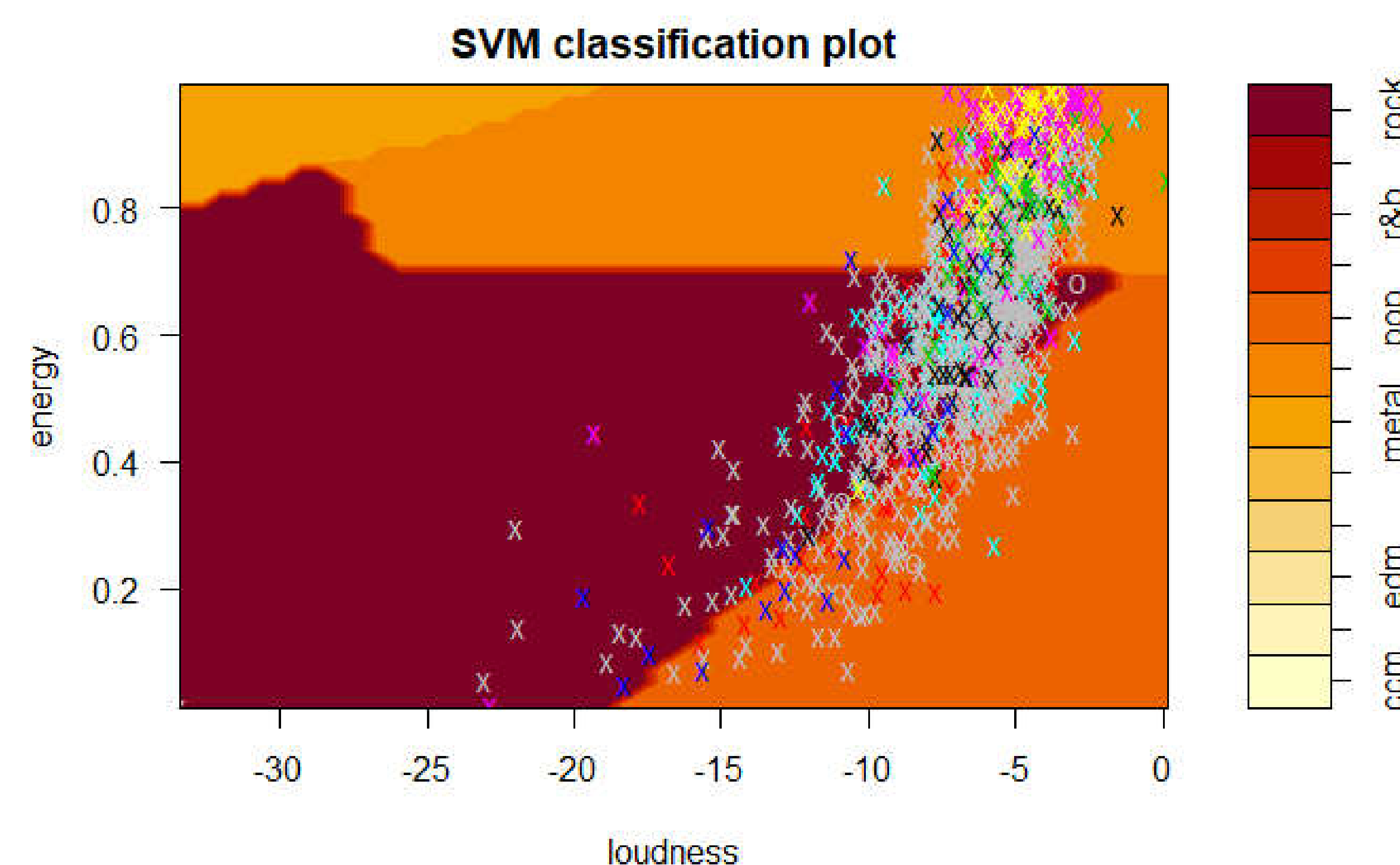
## Classification Tree



**Figure 2: Classification Tree for the determination of Genre**

Genre was determined by key, mode, time signature, acousticness, danceability, energy, instrumentalness, loudness, speechiness, valence, and tempo. The tree has a misclassification rate of about 54%.

## Support Vector Machine



**Figure 3: Linear SVM for Classifying Genre**

A plot of the linear support vector machine showing energy versus loudness. This model has an accuracy of 43%.

|           | Actual |         |     |      |         |       |           |     |      |     |     |      |
|-----------|--------|---------|-----|------|---------|-------|-----------|-----|------|-----|-----|------|
| Predicted | ccm    | country | edm | folk | hip hop | metal | metalcore | pop | punk | r&b | rap | rock |
| ccm       | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| country   | 0      | 10      | 1   | 0    | 0       | 1     | 0         | 5   | 2    | 0   | 0   | 4    |
| edm       | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| folk      | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| hip hop   | 0      | 0       | 0   | 0    | 24      | 0     | 0         | 16  | 0    | 1   | 21  | 3    |
| metal     | 0      | 0       | 1   | 0    | 0       | 23    | 3         | 2   | 2    | 0   | 1   | 9    |
| metalcore | 1      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| pop       | 4      | 60      | 12  | 3    | 28      | 5     | 3         | 159 | 3    | 4   | 28  | 58   |
| punk      | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| r&b       | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| rap       | 0      | 0       | 0   | 0    | 0       | 0     | 0         | 0   | 0    | 0   | 0   | 0    |
| rock      | 3      | 9       | 1   | 2    | 1       | 9     | 3         | 17  | 8    | 0   | 0   | 33   |

**Table 1: Confusion Matrix for SVM Model**

Predicted versus actual for the test dataset