

Billboard Charts



Billboard publishes musical charts of the most popular (highest sales, streams, etc.) songs & artists



Charts are released on a weekly and/or yearly basis

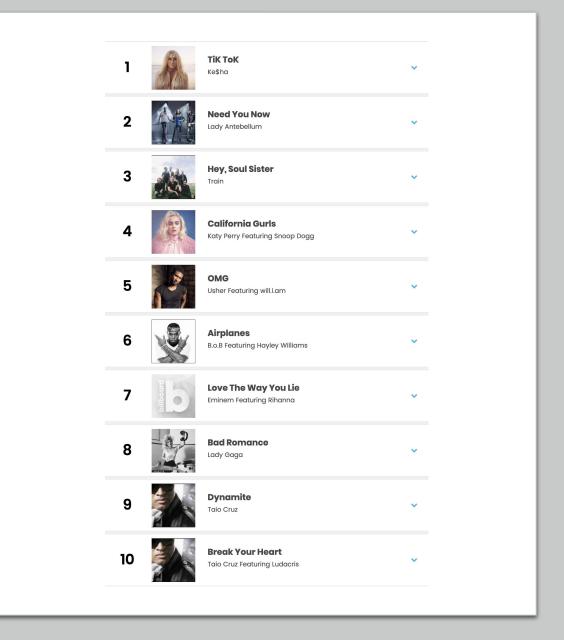


Great way to look at how trends and pop culture changes and evolves

The Hot 100

- The Hot 100 is a popular Billboard chart
- Publishes 100 songs at end of year
- Songs are listed in ascending order— most popular at #1

Top 10 songs in 2010's Hot 100 Chart



The Dataset

Hot 100 songs from 2010-2020 yearly charts

• 1100 songs in total

Split data into 3 sets

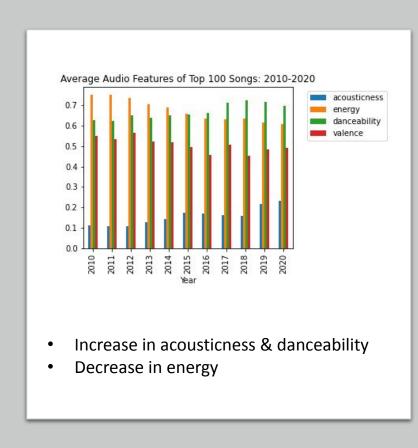
• Train: 2010-2015

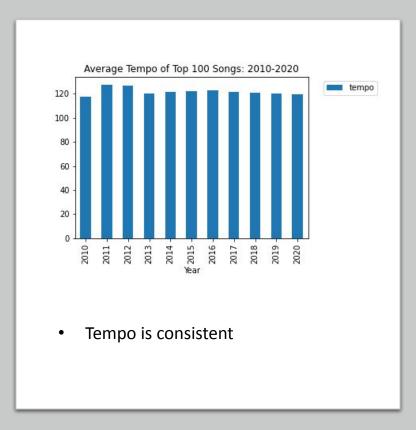
• Validate: 2016-2017

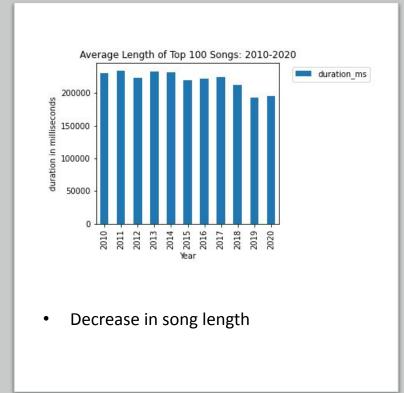
• Test: 2018-2020

Spotify API Features:

- Accousticness
- Danceability
- Duration
- Energy
- Instramentalness
- Key
- Liveness
- Loudness
- Speechiness
- Tempo
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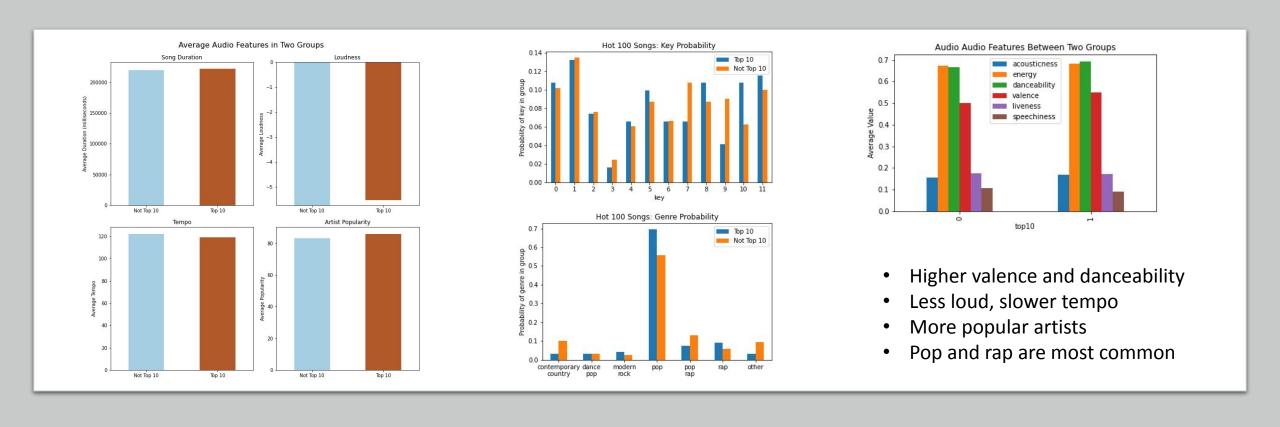


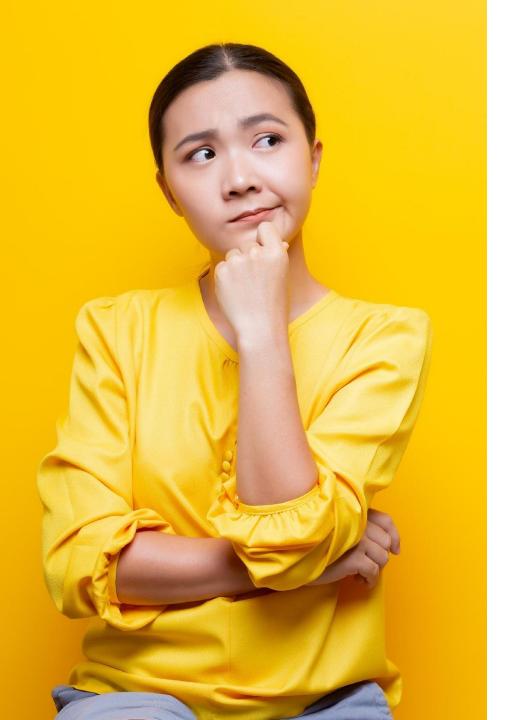




How has the Hot 100 changed over the past decade?

How do **Top 10** songs compare to the rest of the Hot 100?





The Project

Which **features** best predict the **top 10** songs on Billboard Hot 100 charts?

Project Use & Intentions:

- Artists wanting to reach larger audiences can know what to focus on
- Music companies seek out songs that meet these features
- Predicting trends

Project Overview

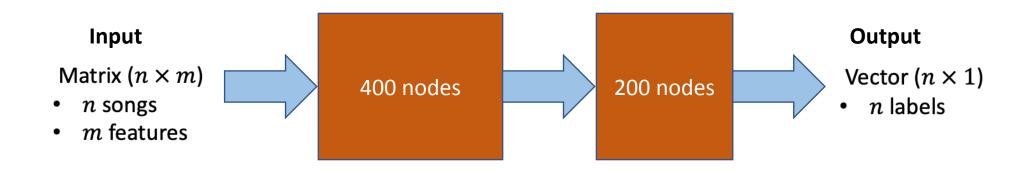
Use neural networks to solve binary classification problem

- Class 0: Not top 10 song (90% of dataset)
- Class 1: Top 10 song (10% of dataset)

Reduce dataset by using feature selection techniques

- Irrelevant features can impact the accuracy of our model
- Which features yield the best predictive model?

Classification with Neural Networks



- 1. Train network with training data
- 2. Use validation data to improve parameters
- 3. Feed network testing data
- 4. Use the network's output and compare with true classes of testing data

Applying Feature Selection to Neural Networks

Train and test the neural network using data that contains:

All features

Features selected by **ExtraTreesClassifier**

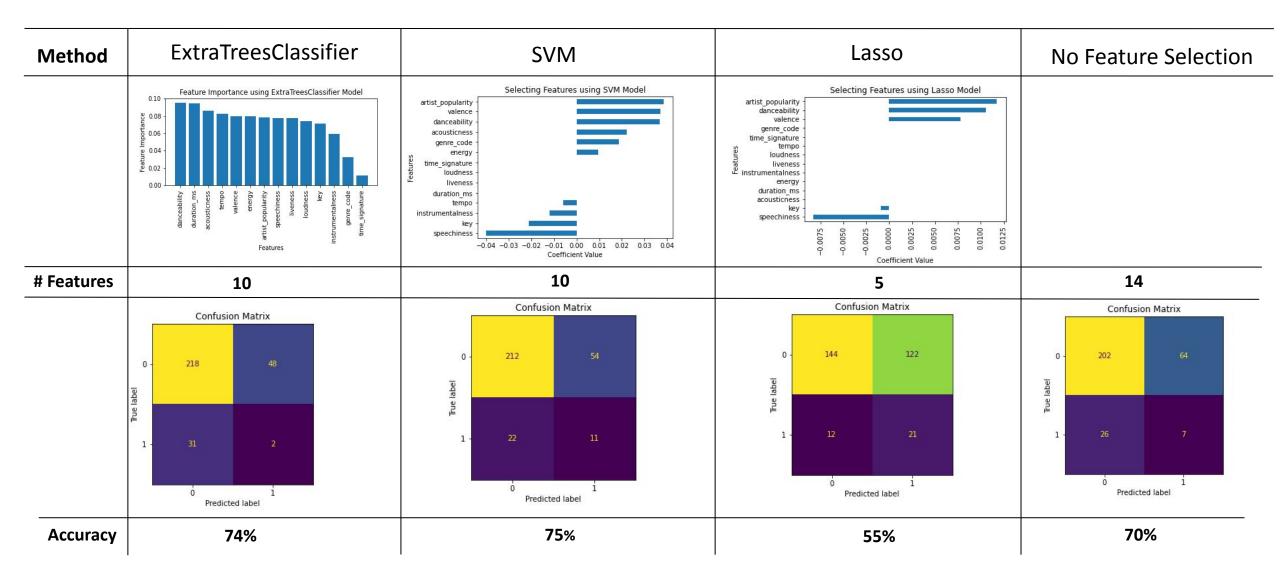
Features selected by Lasso

Features selected by **SVM**

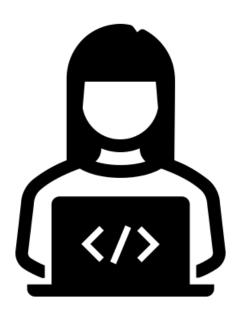


The model with the highest accuracy contains the features most useful with identifying top 10 songs

Results

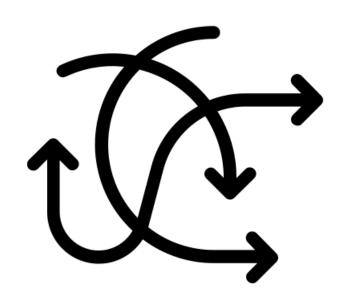


The Team



- I worked alone on this project
- Skills: programming, machine learning, listens to music
- I collected the data and ran the analysis

Project Challenges & Limitations



- Spotify API was irritating
 - Features are only for songs in a playlist or on the same album
 - Had to manually create Spotify playlists for 1100 Billboard songs
- Limited Data
 - Unable to obtain artists net worth demographics
 - No database for this information

Conclusion



SVM performed the best with the following features:

Artist popularity

Danceability

Valence

Genre

Acousticness

Energy

Tempo

Instramentalness

Key

Speechiness



Lasso performed the worst

Had smallest # of features Likely too much data reduction



It'd be interesting to consider additional features

artist net worth artist demographics