# **Module 4 Project**



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The fourth industrial revolution is in full effect, and companies, particularly in the music business, need to prepare a new set of strategies if they are to adapt and take full advantage of Al's wave of change.

# **Top Music Streaming Services**

**Target Audience** 





















lost.fm



### **Problem Definition**

Why do we need context-aware music recommender systems?

Record companies are estimated to invest \$4.5 billion annually worldwide in A&R targeted marketing



On-demand streams in the US for 2018 hit a record high of **534.6 billion total streams** (up 42% from 2017).



Help fans find the best music possible for their specific taste and interest.



Make recommendations smarter because **listeners have limited time** for music consumption.

## **Project Methodology**

#### **Obtain & Scrub**

- Clean, reduce and combine three datasets together
- Create an MVP dataset

#### Model

- Train/Test split
- Fit the model
- Confusion matrix
- Classification report

- Cross validation
- Neural network

Weeks 1-3

### **Explore**

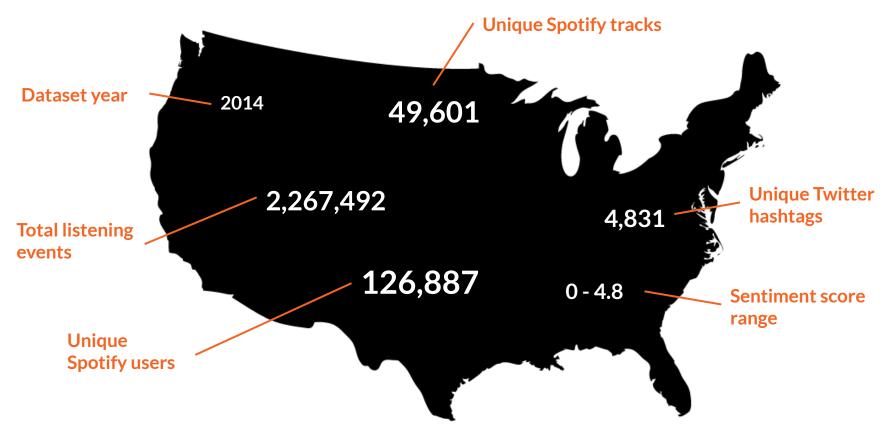
- Check for multicollinearity
- Logistic regression
- Upsample data

Week 4

### Interpret

Sequential neural networks are great at predicting context-aware music recommender systems...

### **About the Data**





#### **By Sentiment Score**

1	thriller	4.8
2	harmonicalove	4.7
3	well	3.9
4	richardmarx	3.6
5	richmond	3.6
6	loversrock	3.4
7	cozypowell	3.4
8	impressionnance	3.3
9	preciousgrace	3.2
10	greatsongforagreatday	3.2

#### By Count/Frequency

1	nowplaying	4,870,436
2	kiss92	56,437
3	postpunk	25,299
4	punk	24,730
5	deathrock	24,699
6	urbantraxxradio	23,867
7	tophits	22,836
8	craveradio	9,735
9	rock	9,483
10	stonerrock	42,97

**Based on #hashtag and time** zone we can use deep learning to predict the next Spotify song you want to listen to!

The model predicts with 99% accuracy.







### Recommendations

Even though the neural network predicts the next song in 81 seconds with 99% accuracy, I recommend using a supervised model for on-demand, music streaming service that can make a prediction in 2 seconds with 99% accuracy.

## **Next Steps / Additional Research**

- Use current Spotify and Twitter datasets (from 2018+) that includes likes/dislikes of songs to determine how additional user context-aware information influences the deep learning, classification model.
- Under a different use case (using current Spotify & Twitter datasets) build a deep learning, classification model based on recommending new music to a user based on their mood, likes/dislikes, and previous listening history.

# THANK YOU

### References

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