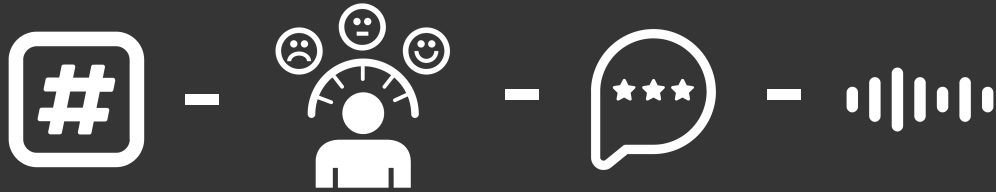


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# Module 4 Project



Chelsea Power

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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 AI's wave of change.

~ Ashley Rose, [Entrepreneur.com](https://www.entrepreneur.com), 2019

# Top Music Streaming Services

Target Audience



# 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 the fan find the best music possible for their specific taste and interest.



Make recommendation and search results 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

Week 4

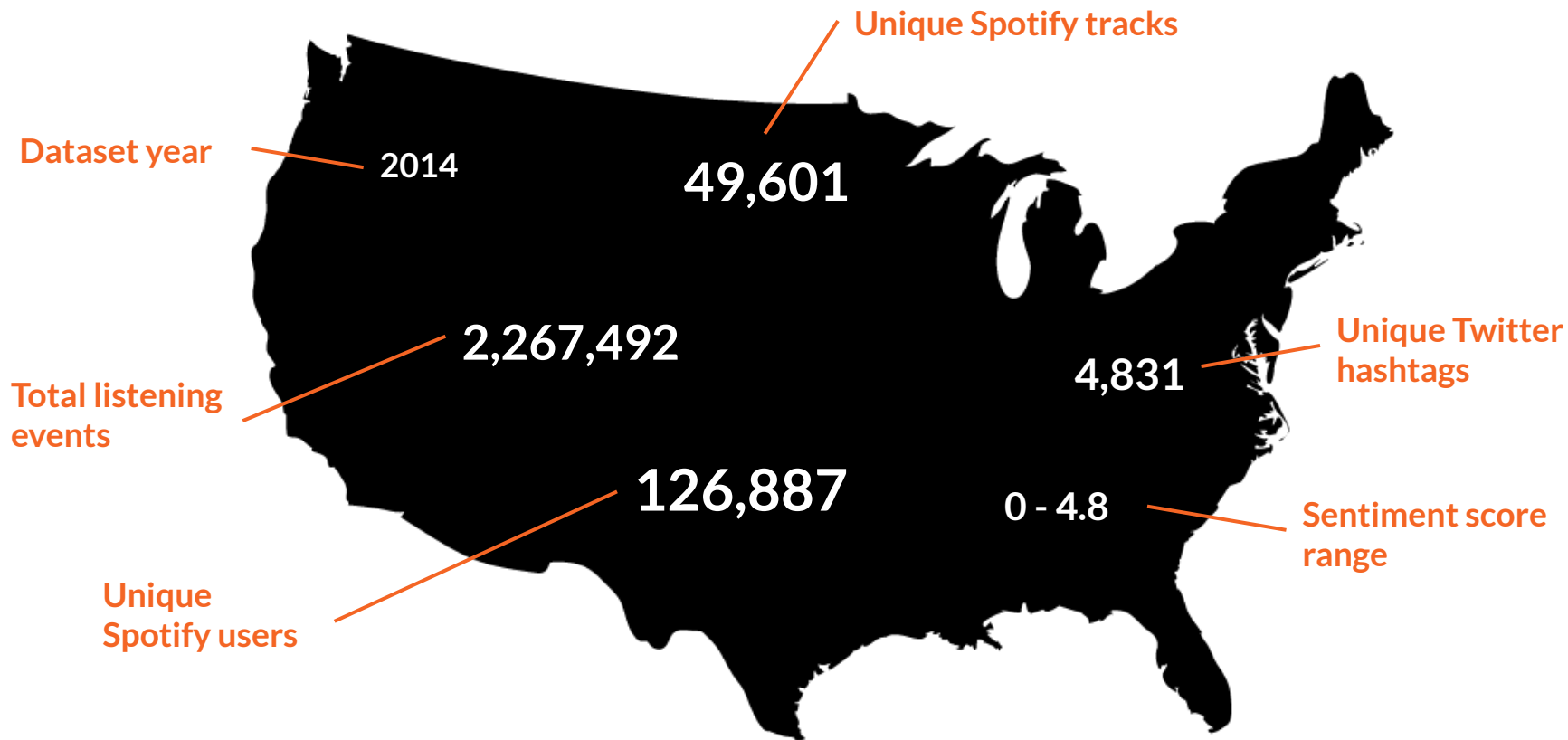
## Explore

- Check for multicollinearity
- Logistic regression
- Upsample data

## Interpret

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

# About the Data



# Top Hashtags

## By Sentiment Score

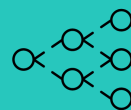
1	thriller	4.8
2	harmonicalove	4.7
3	well	3.9
4	richardmarx	3.6
5	richmond	3.6
6	loversrock	3.4
7	cozypowell	3.4
8	impressionnance	3.3
9	preciousgrace	3.2
10	greatsongforagreatday	3.2

## 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!

My model predicts with **100%** accuracy.





# Next Step Recommendations

- Continue to optimize the neural network so it can predict the next song in **less than 85 seconds**.
- 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.

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# THANK YOU

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# References

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