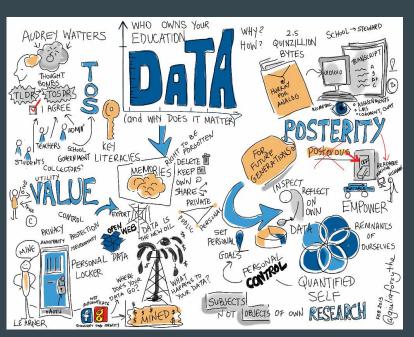
Spotify Music Recommendation Engine

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Data Science Capstone #3







What are Recommendation Engines?

Recommendation Engines are a
 Machine Learning algorithm that
 tries to find patterns in your data

 By learning the patterns about you, Recommendation Engines are able to recommend something to you based on your own personal data



Recommendation Engines are everywhere!

Who uses Recommendation Engines?

- Google
- Amazon
- Facebook
- Spotify / Apple Music
- Netflix
- Youtube
- Any targeted ad on the internet

("cookies")



- A funny cat video
- The perfect pair of socks
- Someone you may know on social media
- The newest movie / video game
- Your future favorite song you have not heard yet

Building my own Recommendation Engine

The Problem:

With the value of recommendation engines almost limitless, I set out to learn more about my the patterns in my own data

The Goal:

To build my own music recommendation engine, and discover some cool, new songs



Data

 The Datasets for this project was all taken from my own personal Spotify account, by using the Spotify API, Spotipy



Datasets broken down into 3 categories:

- "Ratings"
- Top Artist Dataset
- "Train/Test"
- Saved Song Dataset
- "Validation"
- Spotify's Recommended Songs Dataset

EDA

• My Top 10 Artists:

- 1) Young Dolph
- 2) Bo Burnham
- 3) Laurent Korcia
- 4) Black Eyed Peas
- 5) The Beatles
- 6) Oasis
- 7) Drake
- 8) Key Glock
- 9) Gucci Mane
- 10) Kanye West





My Top 10 Songs:

- 1) Les Valseuses
- 2) Yebba's Heartbreak
- 3) Came For The Low
- 4) READY TO GO
- 5) Don't Wanna Know
- 6) SUVs (Black on Black)
- 7) Who Want Smoke??
- 8) JU\$T
- 9) Garden Party
- 10) Still Trappin'

Model Metrics

Classification Problem

 Used the "top artist" dataset to rate the "training/test" dataset. "1" rating if the song was by one of my top artists, "0" if not

- Trained data with song content variables, such as:
- Tempo
- Loudness
- Danceability
- Accousticness
- Key
- Liveness

- Valence
- Speechiness
- Energy
- Instrumentalness
- Mode
- Genre

Applied rating to 399 songs in a dataset of 1133

Model Metrics

• My Top 10 features of importance:

1. rap: 0.0733806

2. speechiness: 0.05040762

3. energy: 0.0453377

4. danceability: 0.04341639

5. instrumentalness: 0.04269710

6. loudness: 0.0409288

7. acousticness: 0.0350782

8. detroit hip hop: 0.03430248

9. tempo: 0.0318333

10. valence: 0.0313792



Modeling

- Fit 3 Different types of
 Classification Models and
 compared accuracy to determine
 best model:
 - Logistic Regression
 - o kNN
 - o Random Forest Classifier

Accuracy Results:

- <u>Logistic Regression</u>: 0.665
- <u>kNN</u>: 0.656
- Random Forest Classifier: 0.947

Recommendation Engine Outcomes

By using my trained models with my top artist spotify data, I was able to successfully predict 116 songs, both old and new, that I would enjoy with a probability of over 0.75



(Me after listening the the recommended songs)

Recommendations for how to use

- Model can be trained specific to anyone, by using other Spotify users data
- Simply enter in personal data gathered from the Spotipy API, and my model can recommend songs specific to you!
- Datasets can be "rated" using other methods, such as: favorite songs



Conclusions

Song Recommendations were lit!





 Because I used "top artist" data, and my top artists were predominately rap/hip hop genre, and the recommendations with the highest probability were all of a similar genre/ style

 Model could be more accurate if given a more complex rating system, such as rating every song specifically, etc



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