

A linear regression project by JG Premkanna

#### **Process**

Step 1 Step 2 Step 3

- Gather 2,000 song data from Kaggle and additional data from Spotify API
- Clean data for modeling

Use EDA & Statistical Testing to gain insight into the dataset:

- How my target and explanatory variables relate
- The relationship between x variables

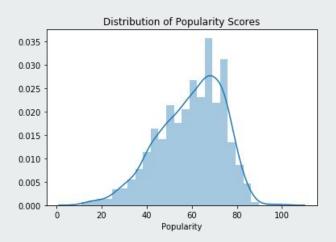
Find the best model for my dataset through:

- Feature transformation
- Feature engineering
- Feature selection
- Compare/contrast model iterations

### **The Data**

### **Target Variable**

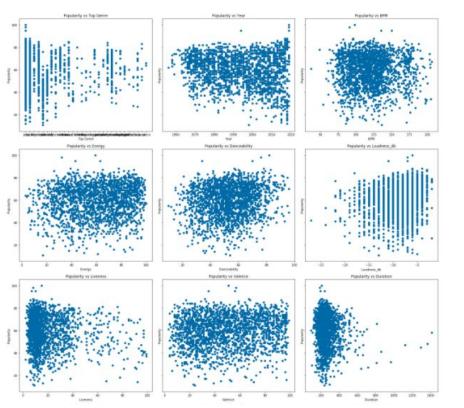
Popularity Score

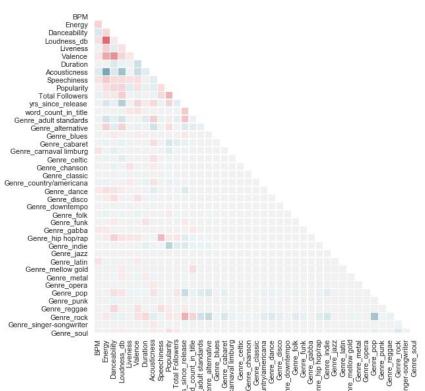


#### **Independent Variables**

- Artist follower count
- Top genre
- # of years since released
- # of words in title
- Audio elements:
  - o BPM
  - Energy
  - Danceability
  - Loudness
  - Valence
  - Speechiness

## **EDA Conclusions**





# **Feature Engineering**

- Use of Polynomial transformations to create interactions such as:
  - Energy Danceability
  - o BPM Loudness
  - Accoustioness Speechiness
- Interaction features create a more predictive model

# **Models & Results**

	# of Features	Train RMSE	Test RMSE	R <sup>2</sup>
Baseline	1	13.63	13.12	.096
All Features	85	11.67	11.83	.26
Select Features	34	11.86	11.69	.28



# Biggest increasers of song popularity:

- # of years since release
- Artist Followers
- Danceability

# Biggest decreasers of song popularity:

- Indie Genre
- AcousticnessSpeechiness

# Final Thoughts

- Only 28% of the amount of variation in song popularity is explained by my final model.
- To increase the predictive power of my model, I would like to try further degrees of polynomial transformations to find better interactions.
- It is also worth exploring other types of models that would be better suited to this dataset.