

March 30, 2022



**SPOTIFY: SONG ANALYSIS**

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BY: SYDNEY TRAN

# Project Statement

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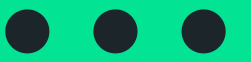
#### Project Statement



Utilizing Spotify's Top 200 Weekly (Global) charts from 2020 and 2021, can we identify what makes a song popular based on specific features of a song?

**Model performance will be determined by the RMSE and R2 score. The success of the model will be measured by an increase of at least 10% from the baseline score.**

# Project Roadmap



## DATA CLEANING & EDA

- Genres
- Outliers

## PREPROCESSING

- Polynomial Features
- One-Hot Encoding  
Categorical Features

## MODELING

- Linear Regression/LASSO
- Decision Trees
- Random Forest

## RMSE

Measure of error (in terms of popularity score 0-100)

## R<sup>2</sup>

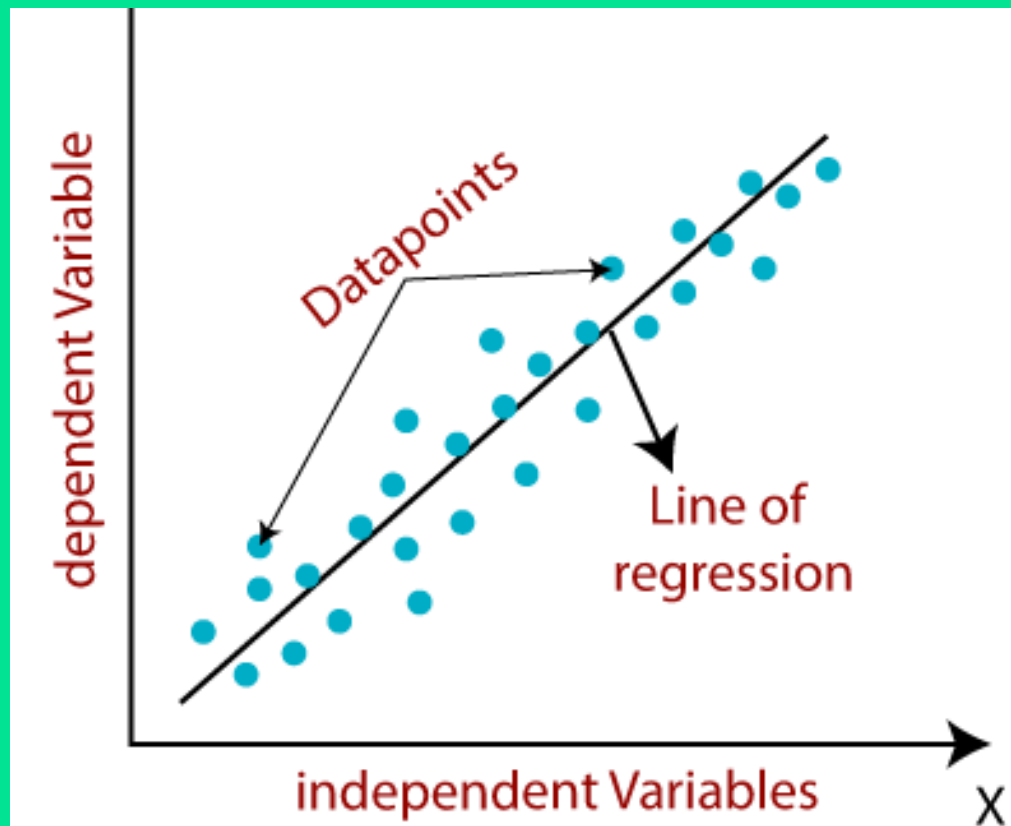
"Statistical measure that represents the proportion of the variance for a dependent variable that's explained by an independent variable"

## POLYNOMIAL FEATURES

Features that are created by raising existing features to an exponent

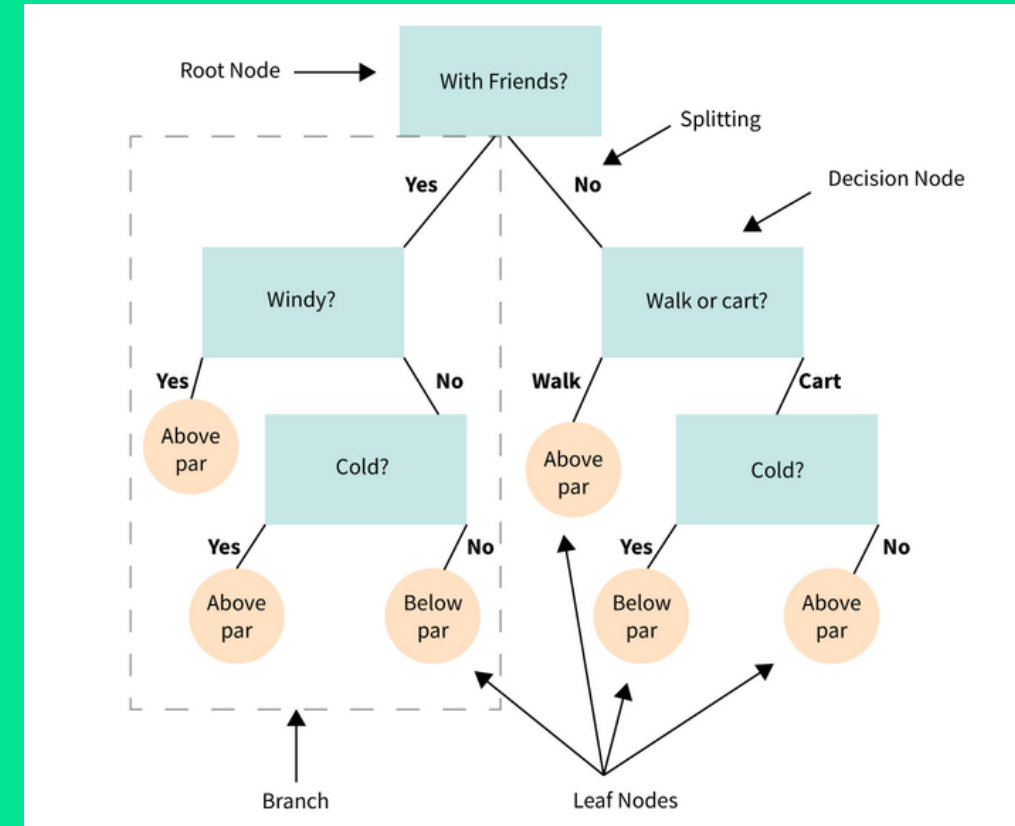
## ONE-HOT ENCODING

Converting categorical data into binary representation



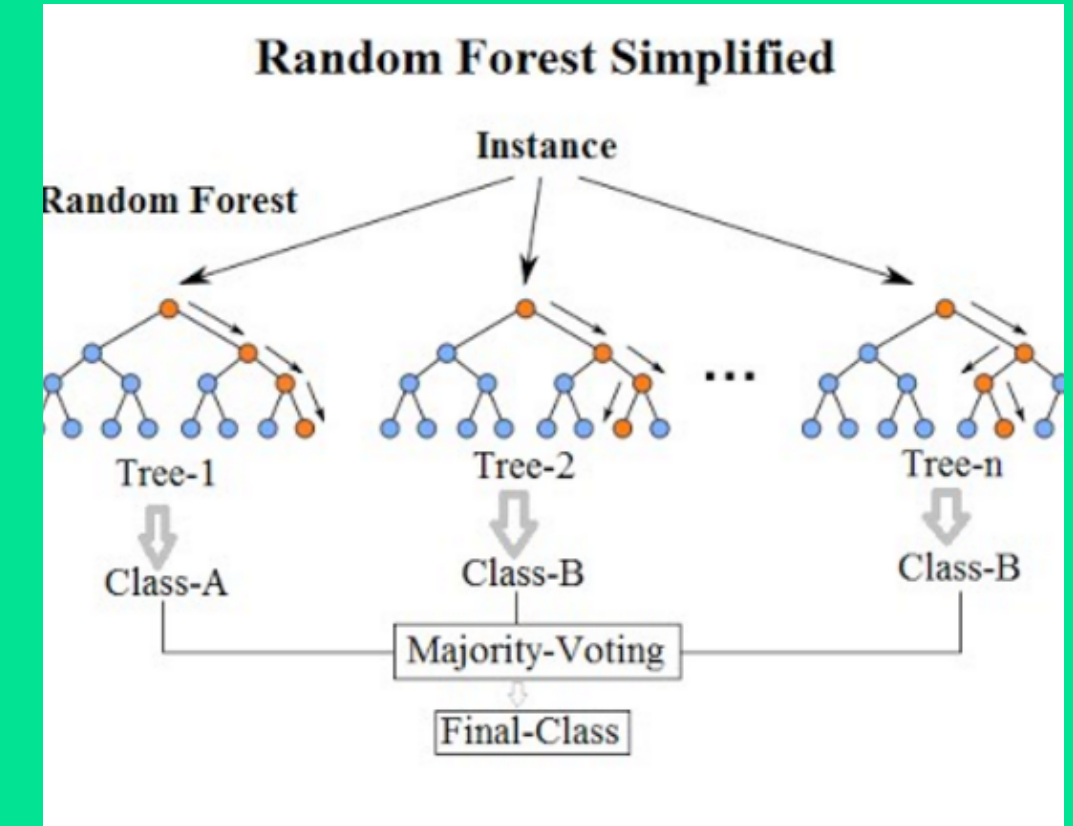
## LINEAR REGRESSION

Establishing a relationship between the dependent and independent variable



## DECISION TREES

Takes a dataset, finds rules based on the X data and splits data into smaller datasets



## RANDOM FOREST

A number of decision trees on various subsamples of datasets



## Your top genres were

#1  
Pop

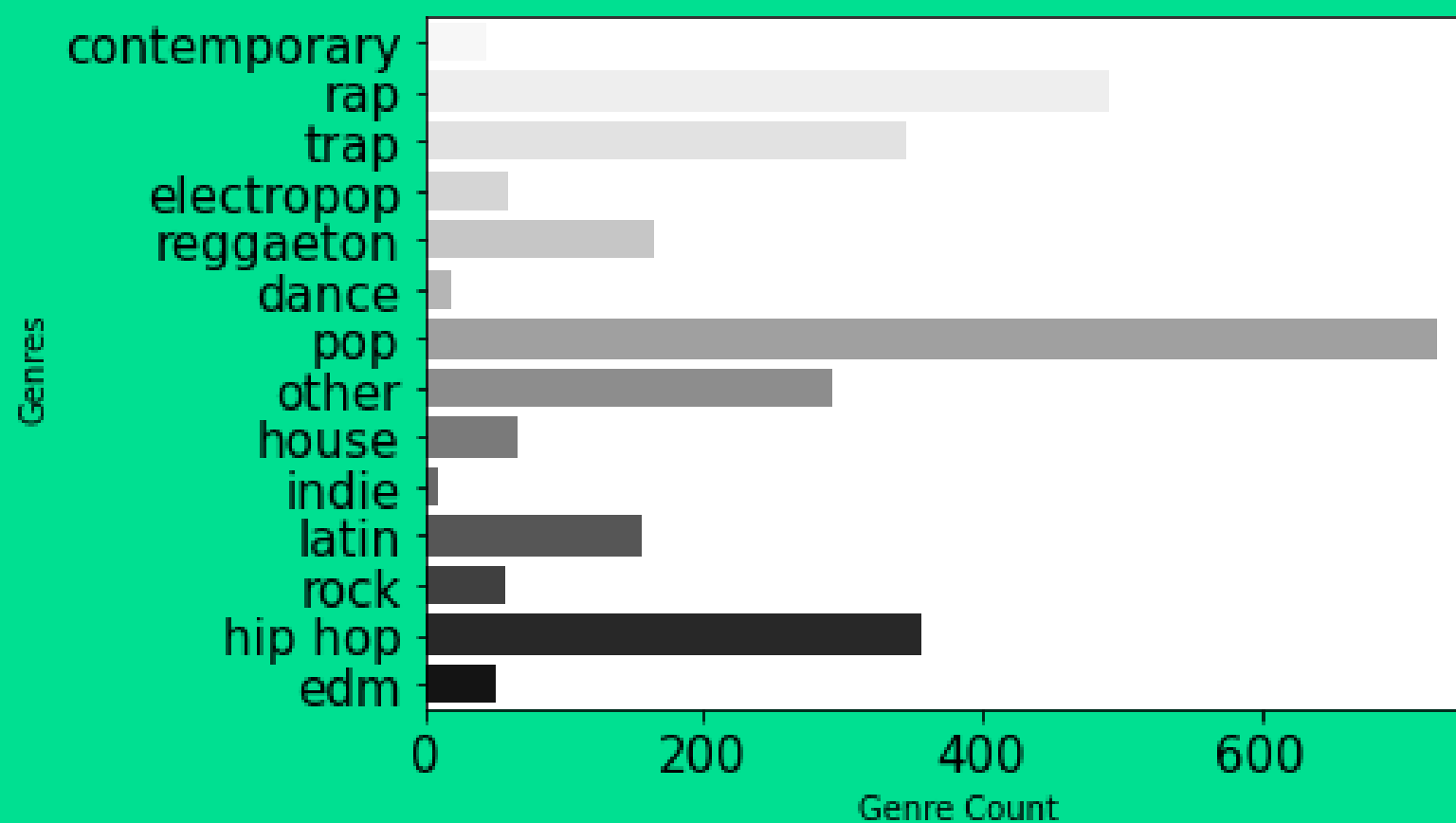
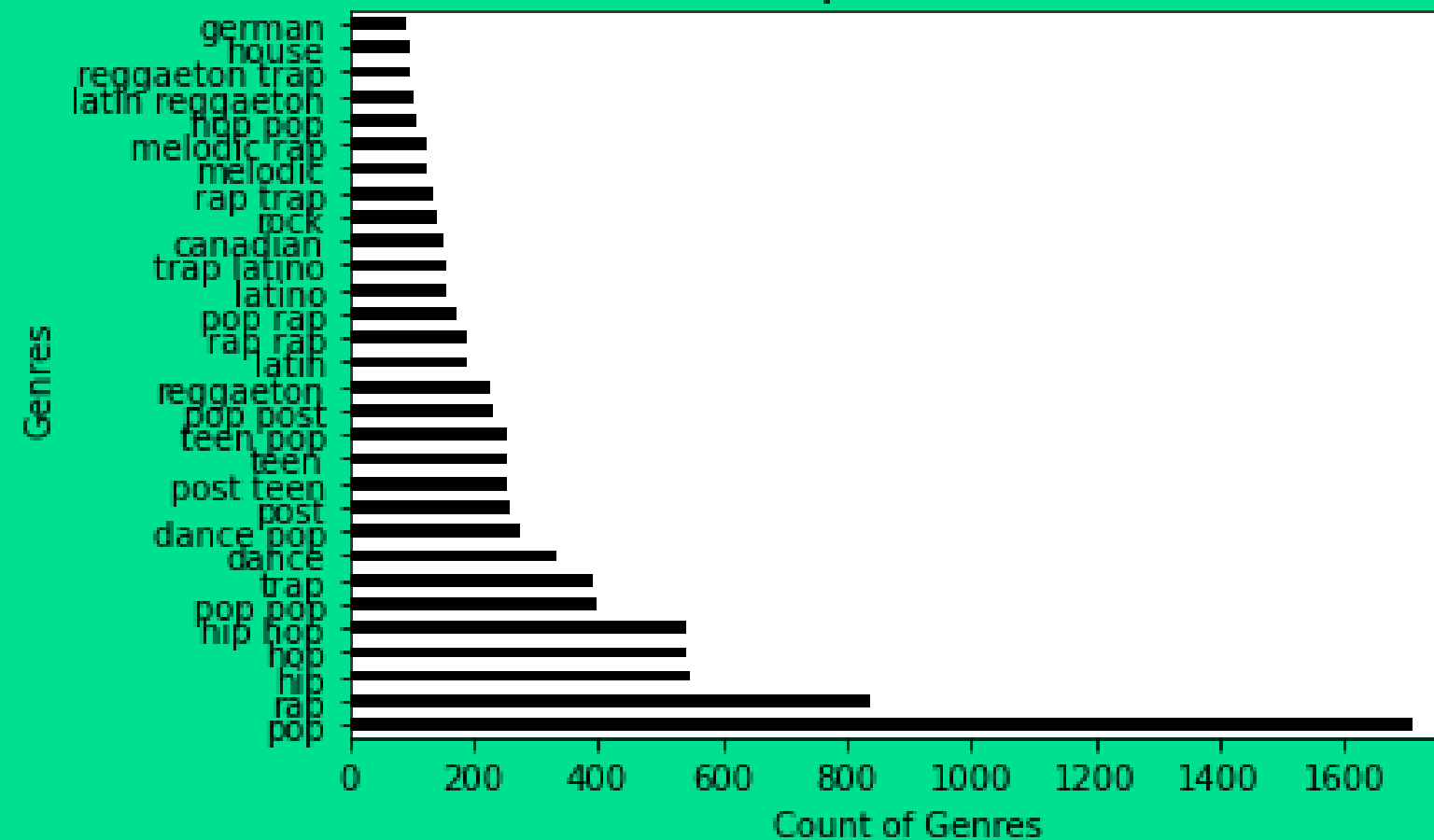
#2  
Rap

#3  
Hip Hop

#4  
Trap

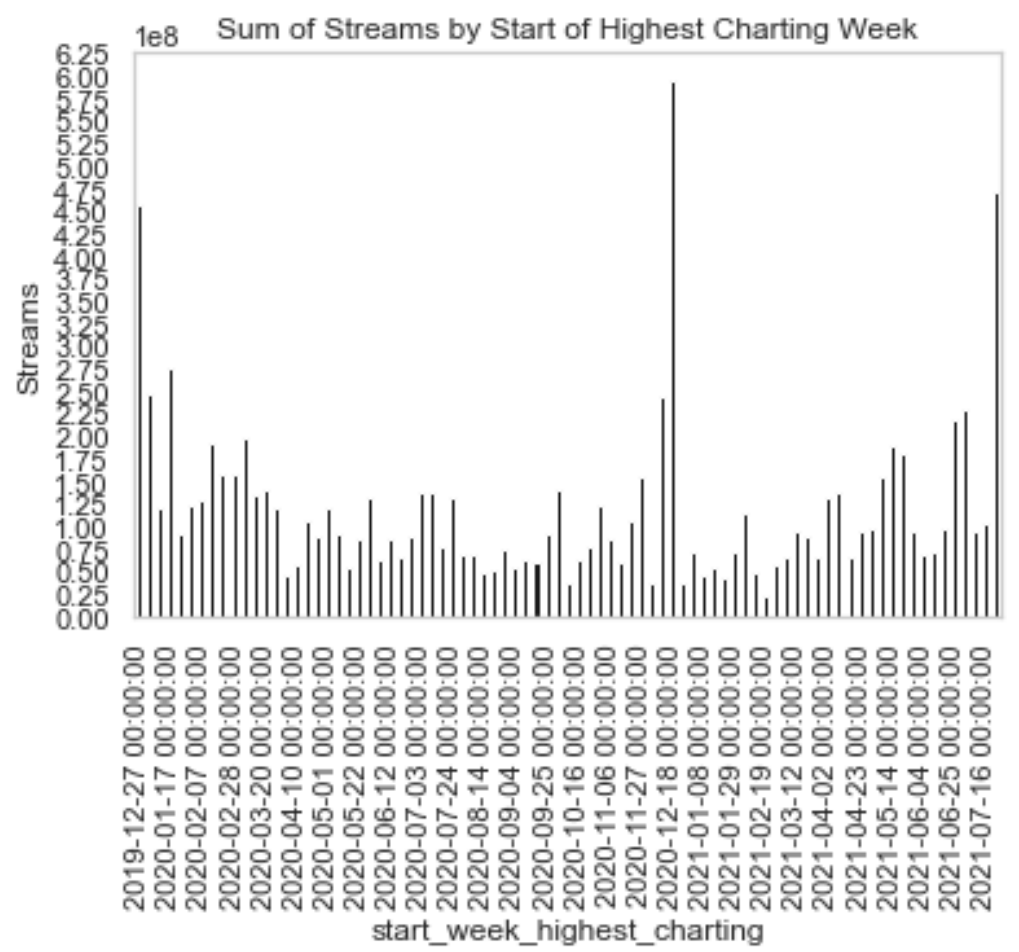
#5  
Dance

## Top Genres



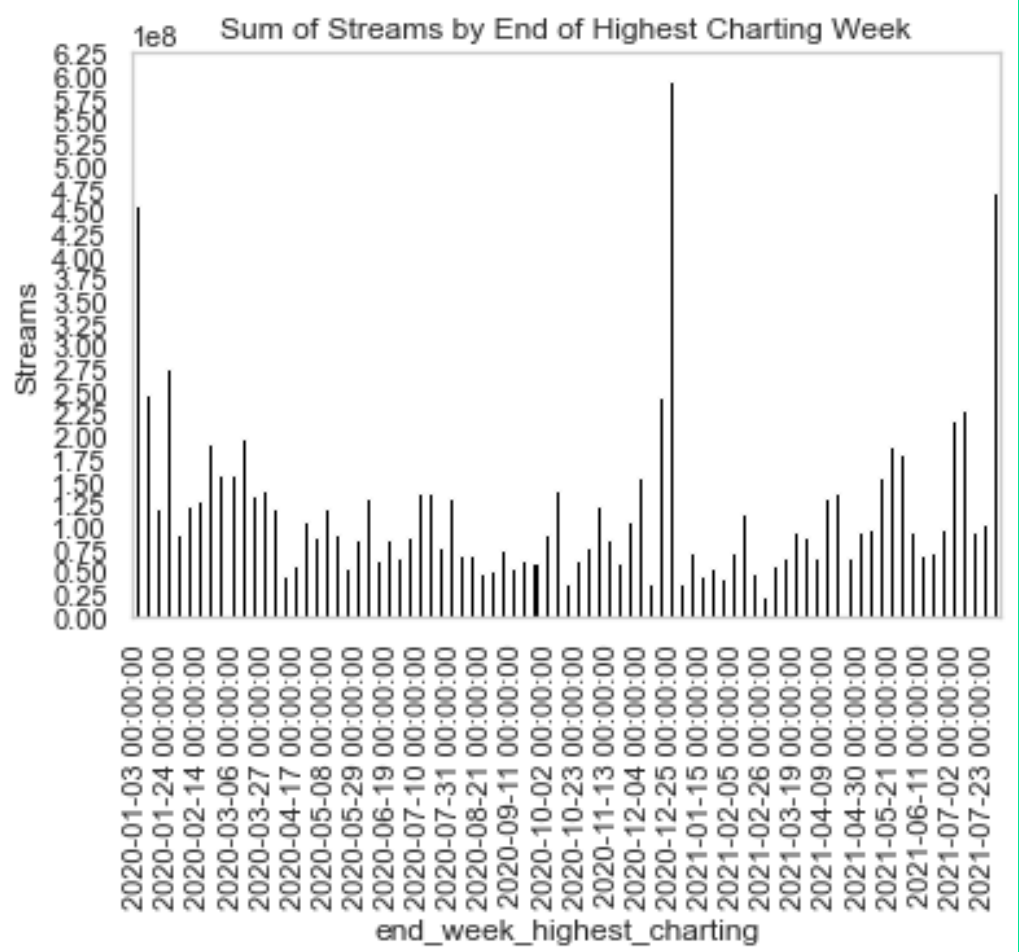
**Lists of 335 various genres to  
14 larger "family" genres**

# Christmas Songs: Outliers



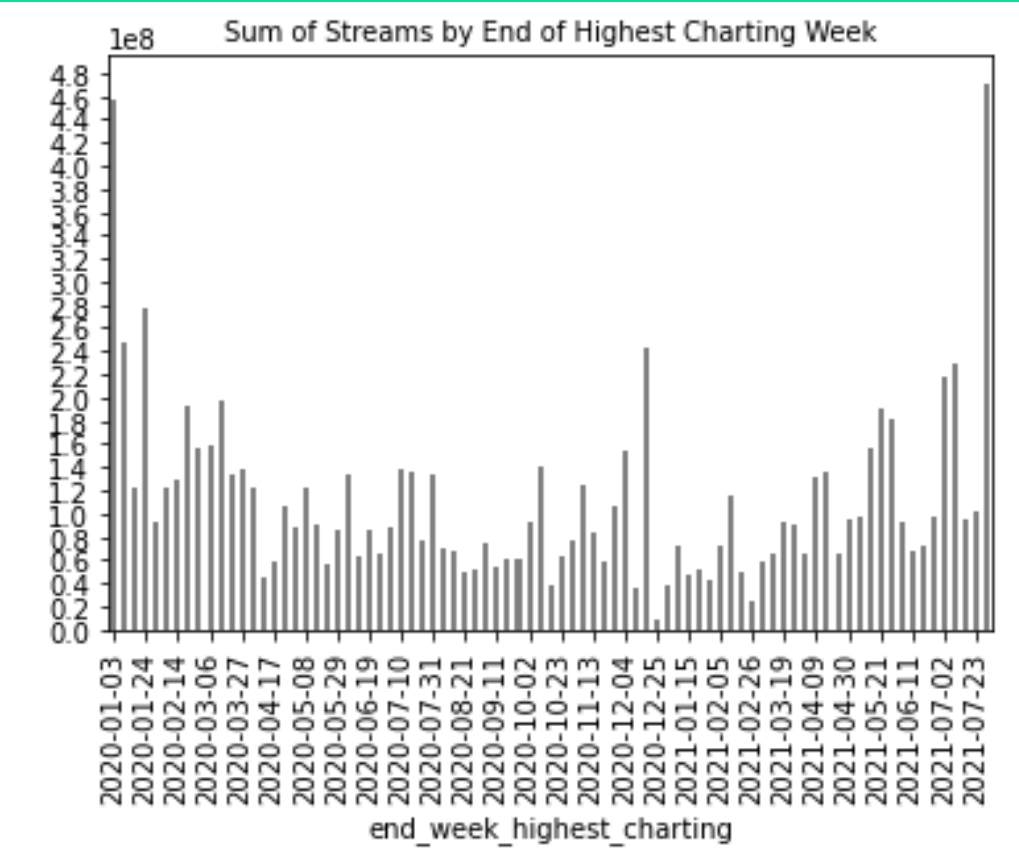
START OF HIGHEST  
CHARTING WEEK

2021-12-18



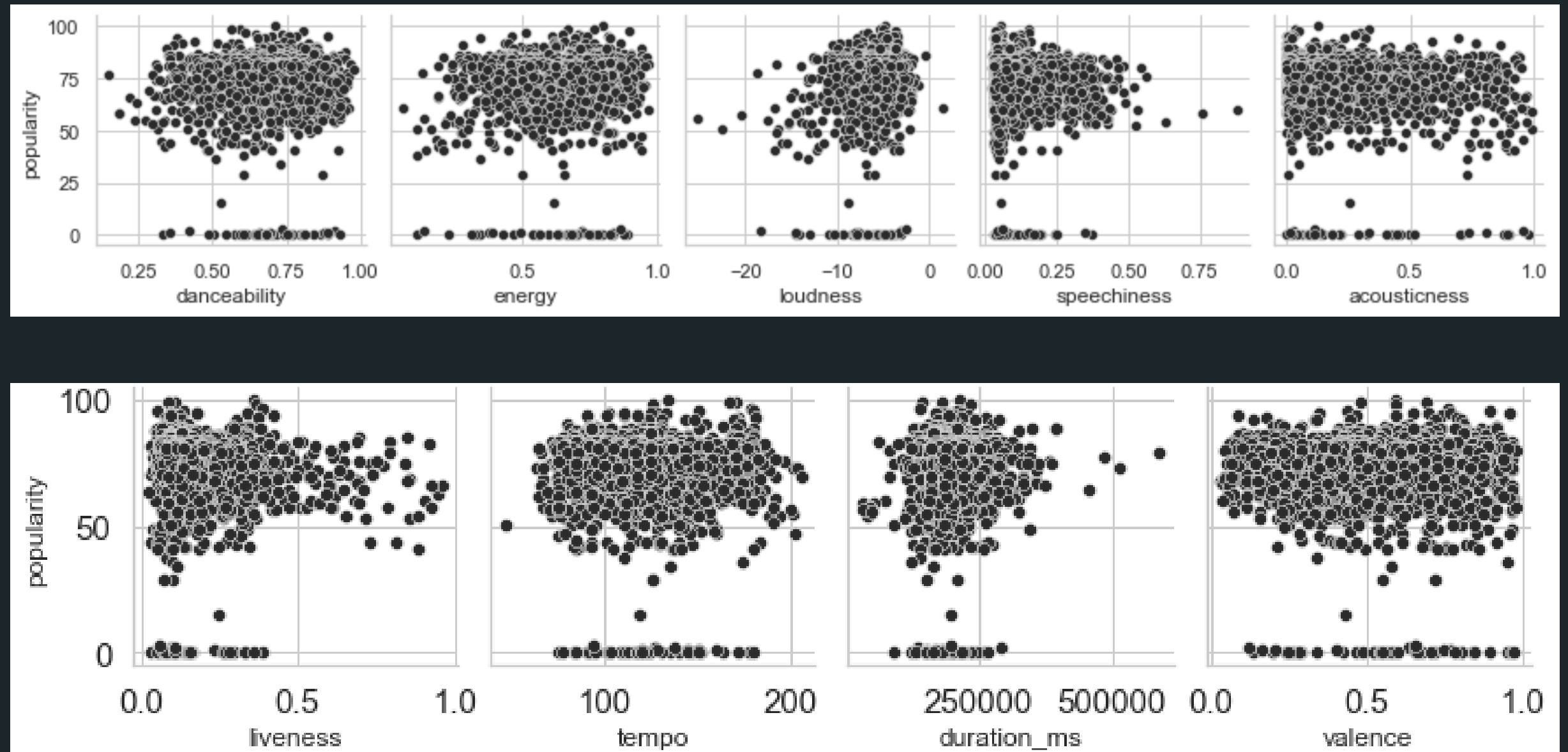
END OF HIGHEST  
CHARTING WEEK

2021-12-25



OUTLIERS  
REMOVED

# CORRELATIONS: POPULARITY



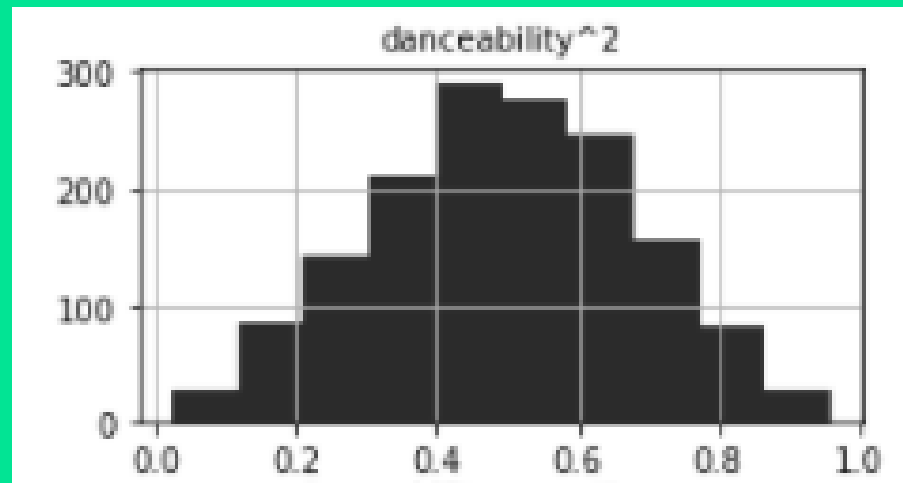
No clear correlations between popularity  
and numerical features



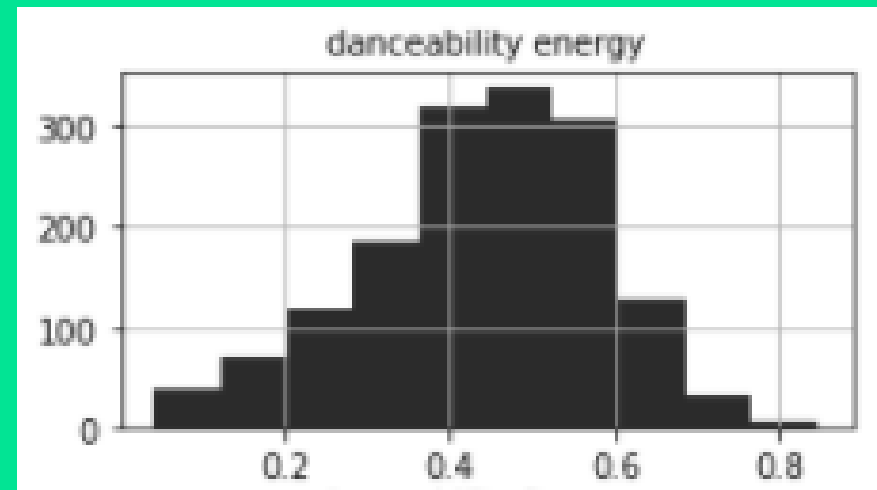


# Polynomial Features

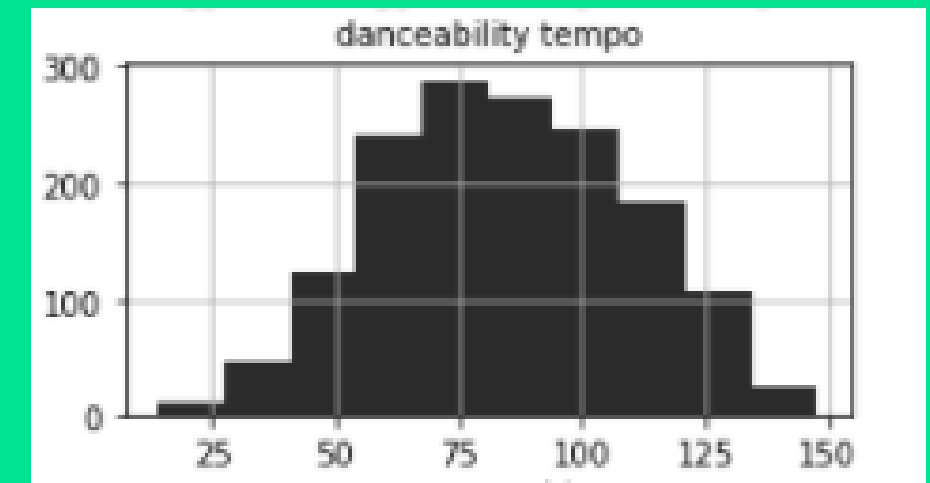
DANCEABILITY<sup>2</sup>



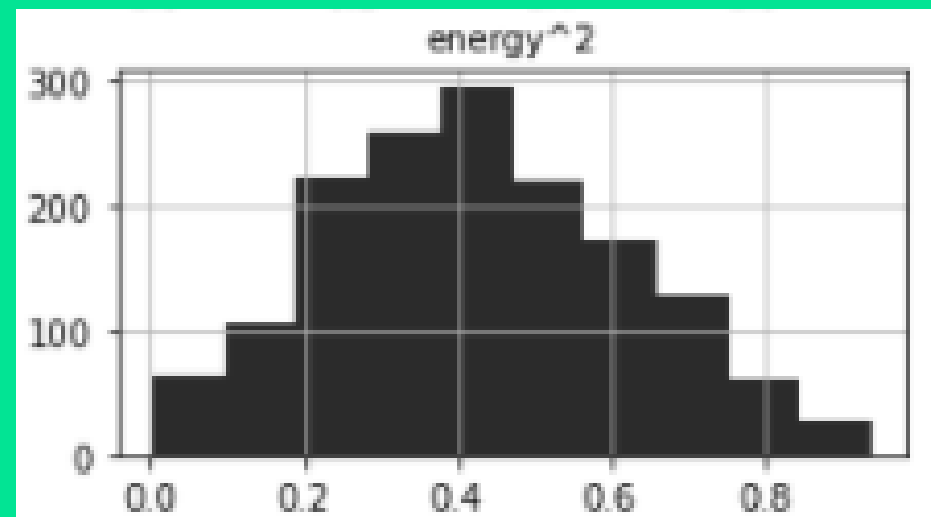
DANCEABILITY  
ENERGY



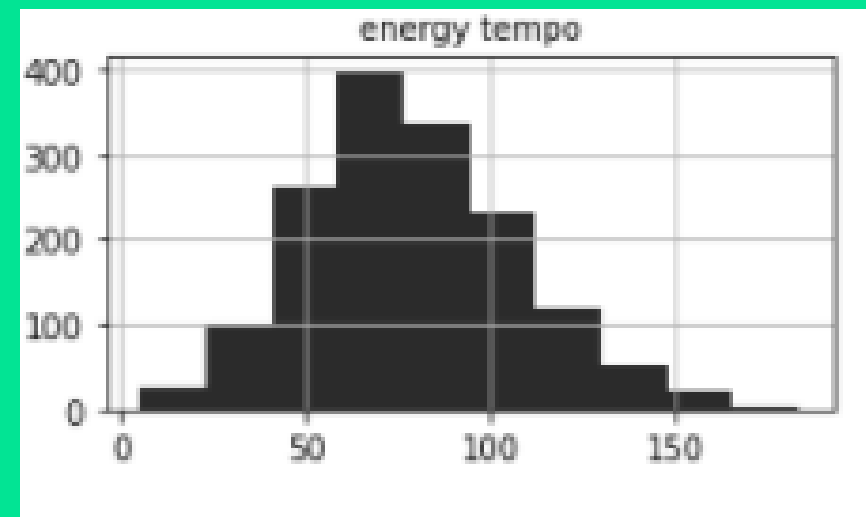
DANCEABILITY  
TEMPO



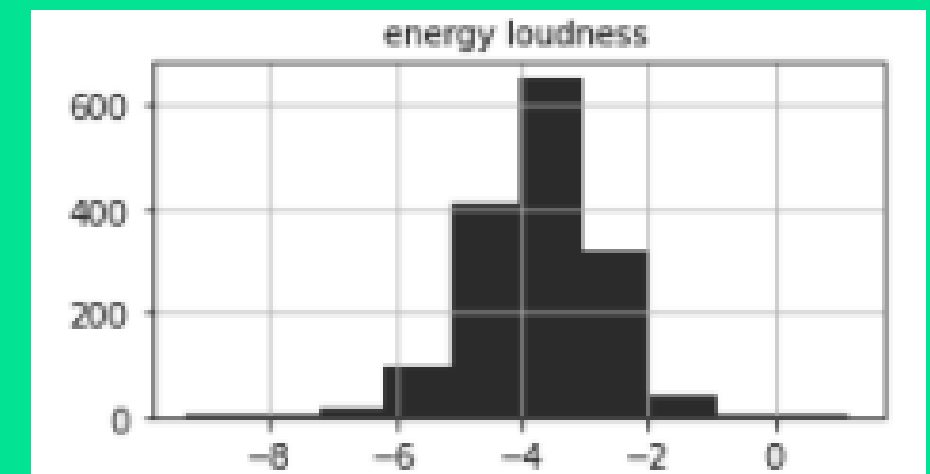
ENERGY<sup>2</sup>



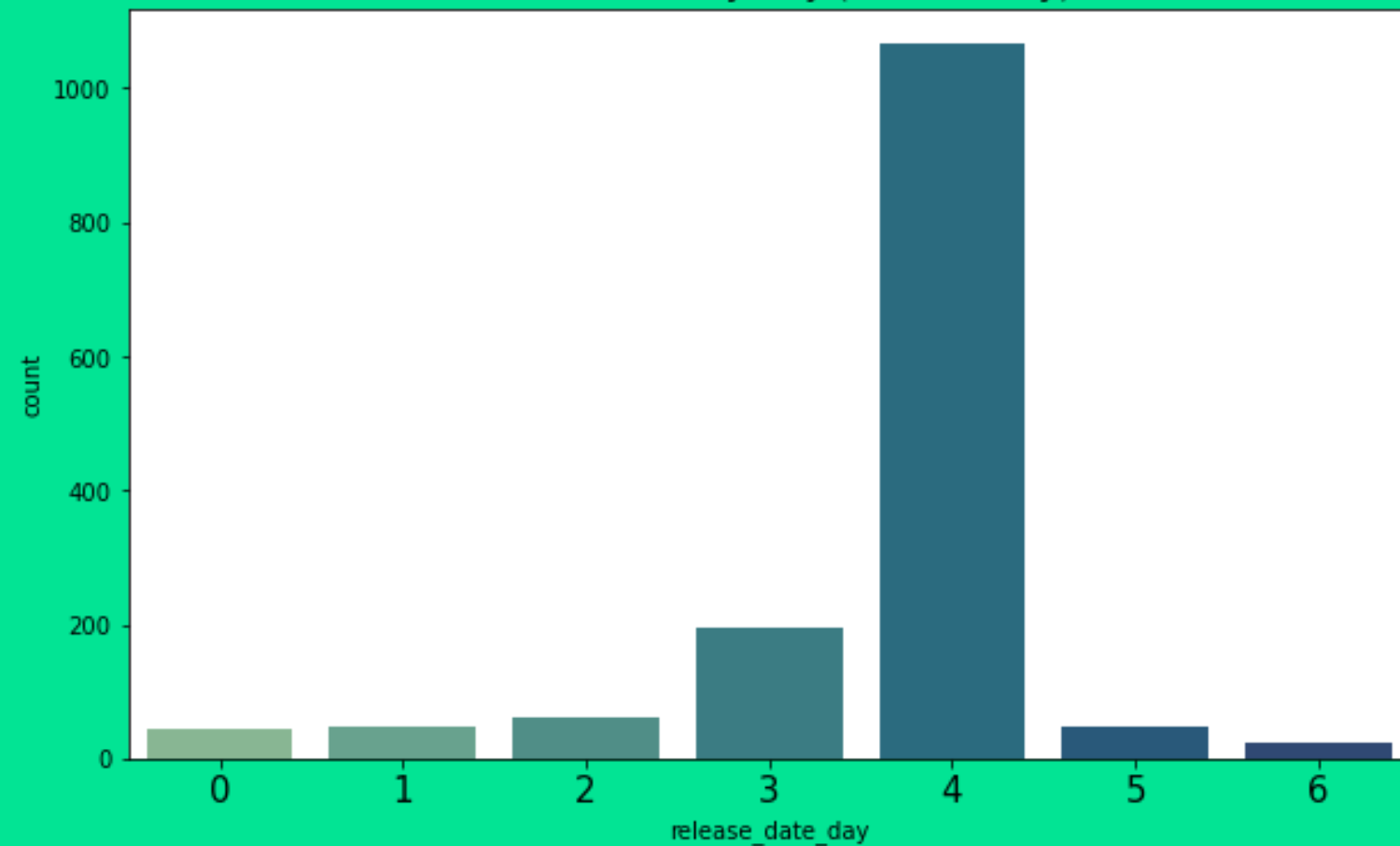
ENERGY TEMPO



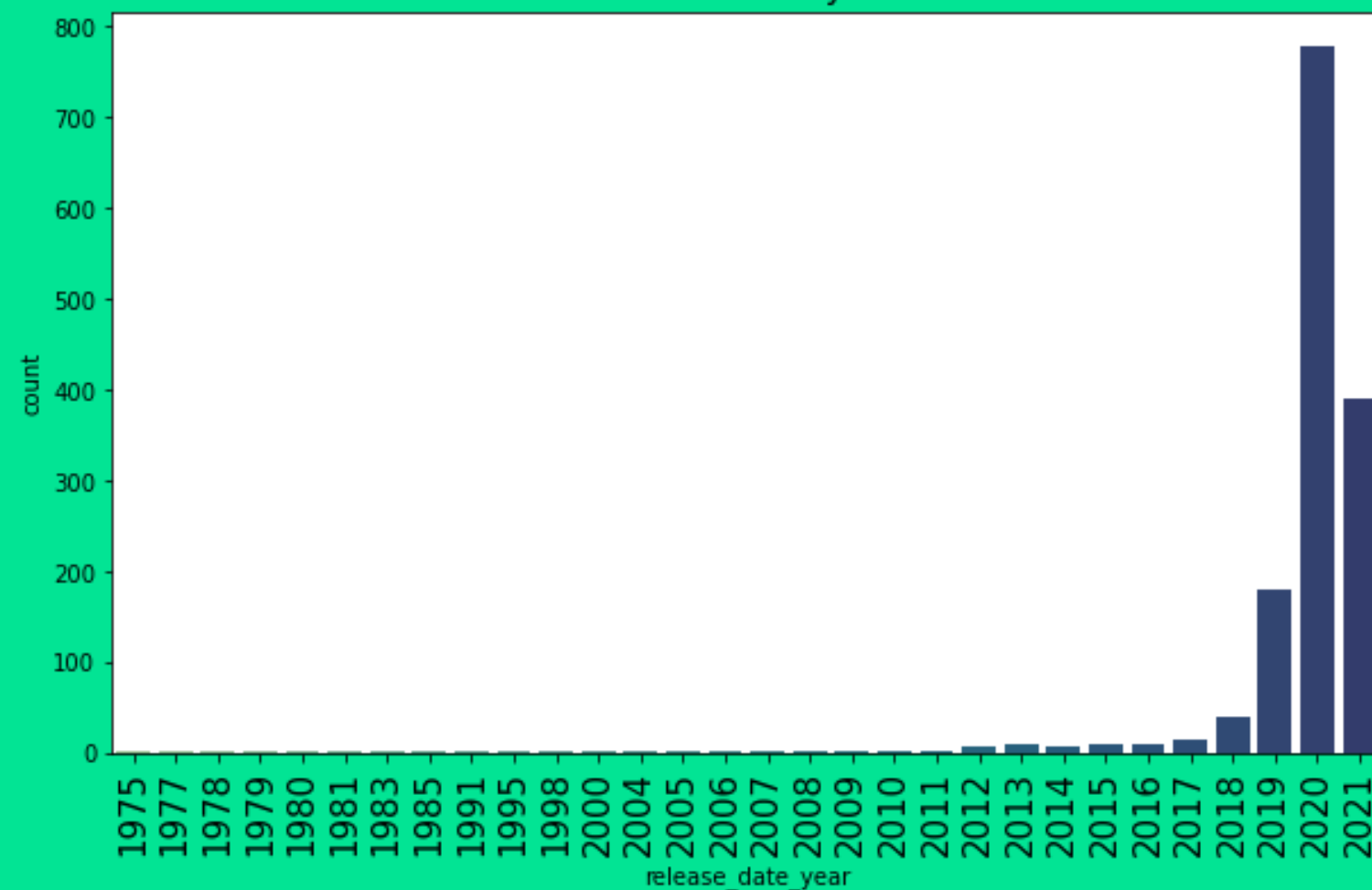
ENERGY  
LOUDNESS



Release Date by Day (0 = Monday)



Release Date by Year



# Release Date

...

**CREATED 2 NEW FEATURES:  
DAY & YEAR**

# Baseline Models

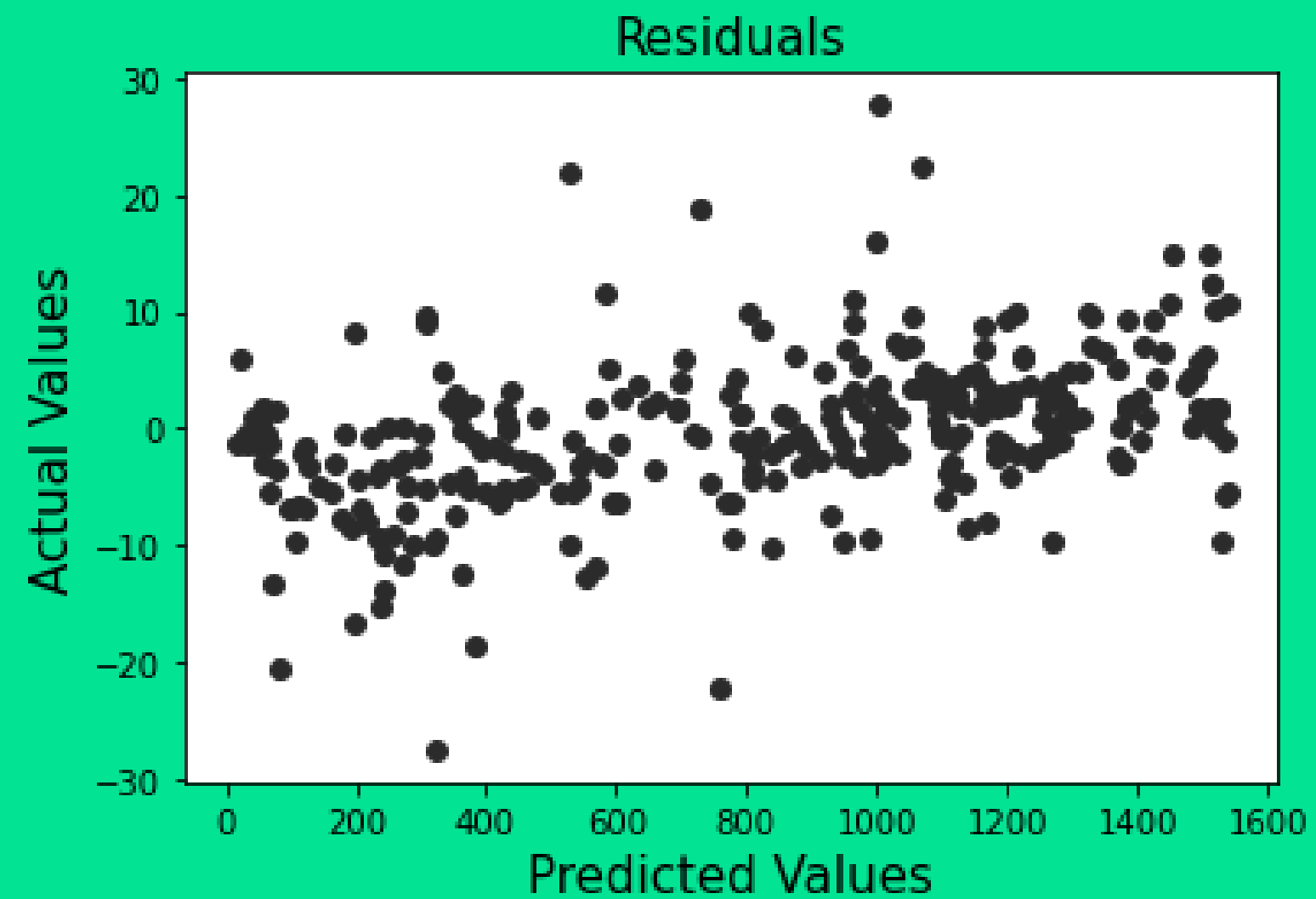
## R2

Model	Train Score	Test Score
Baseline: Linear Regression	0.159	0.119
Baseline: Decision Tree	1.0	0.380
Baseline: Random Forest	0.951	0.744
Decision Tree: Gridsearch	0.725	0.711
Random Forest: Gridsearch	0.923	0.755

## RMSE

Model	Train Score	Test Score
Baseline: Linear Regression	14.37	15.39
Baseline: Decision Tree	0.0	12.91
Baseline: Random Forest	3.48	8.29
Decision Tree: Gridsearch	8.22	8.82
Random Forest: Gridsearch	4.35	8.11

polynomial features  
polynomial features  
polynomial features  
polynomial features



## RANDOM FOREST – POLYNOMIAL FEATURES ADDED

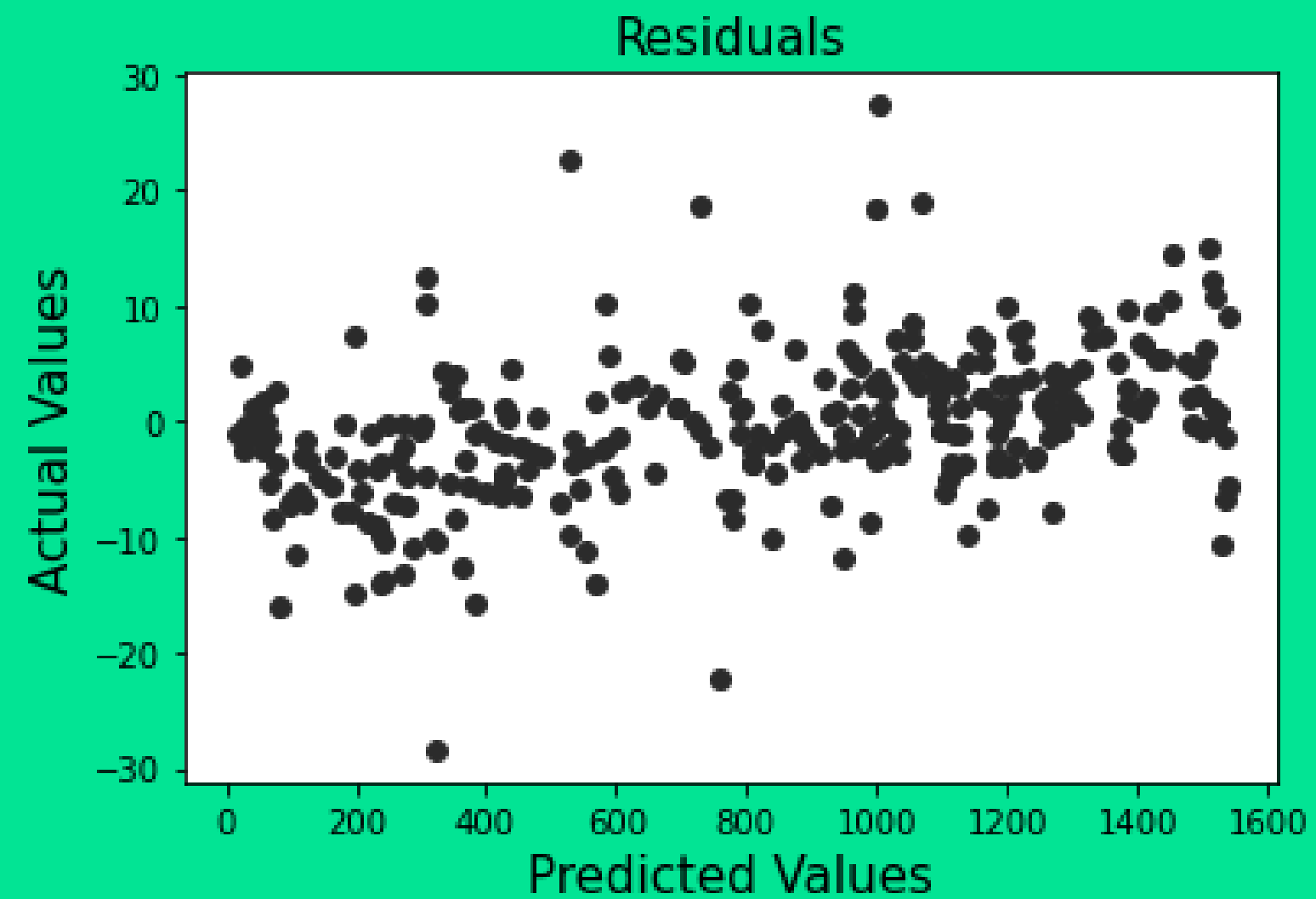
R2 Train Score: 0.962

RMSE Train Score: 2.91

R2 Test Score: 0.794

RMSE Test Score: 6.66

chord  
chord  
chord  
chord



## RANDOM FOREST – CHORD ADDED

R2 Train Score: 0.962

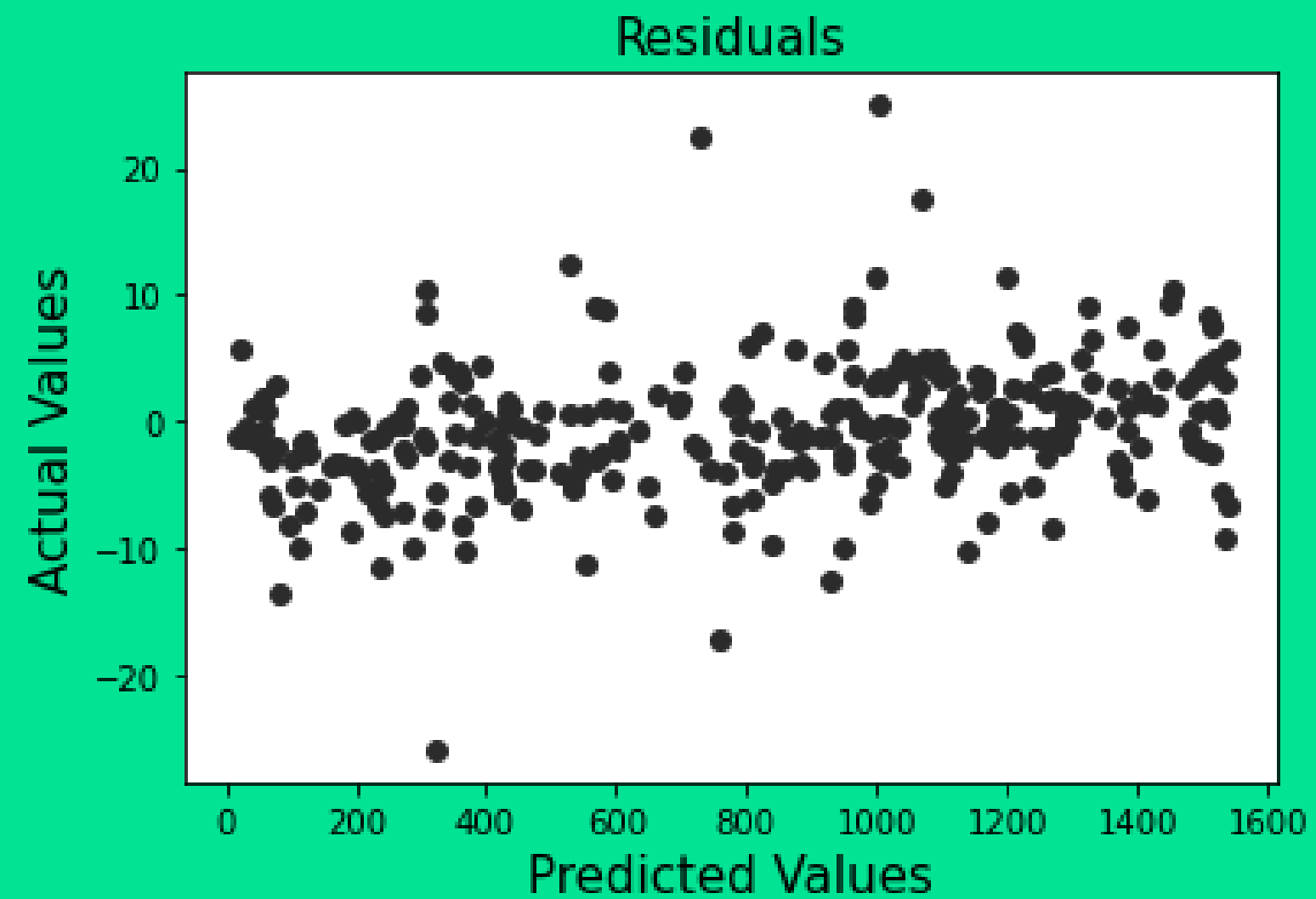
RMSE Train Score: 2.93

R2 Test Score: 0.799

RMSE Test Score: 6.58



release date & year  
release date & year  
release date & year  
release date & year



## RANDOM FOREST – RELEASE DATE & YEAR ADDED

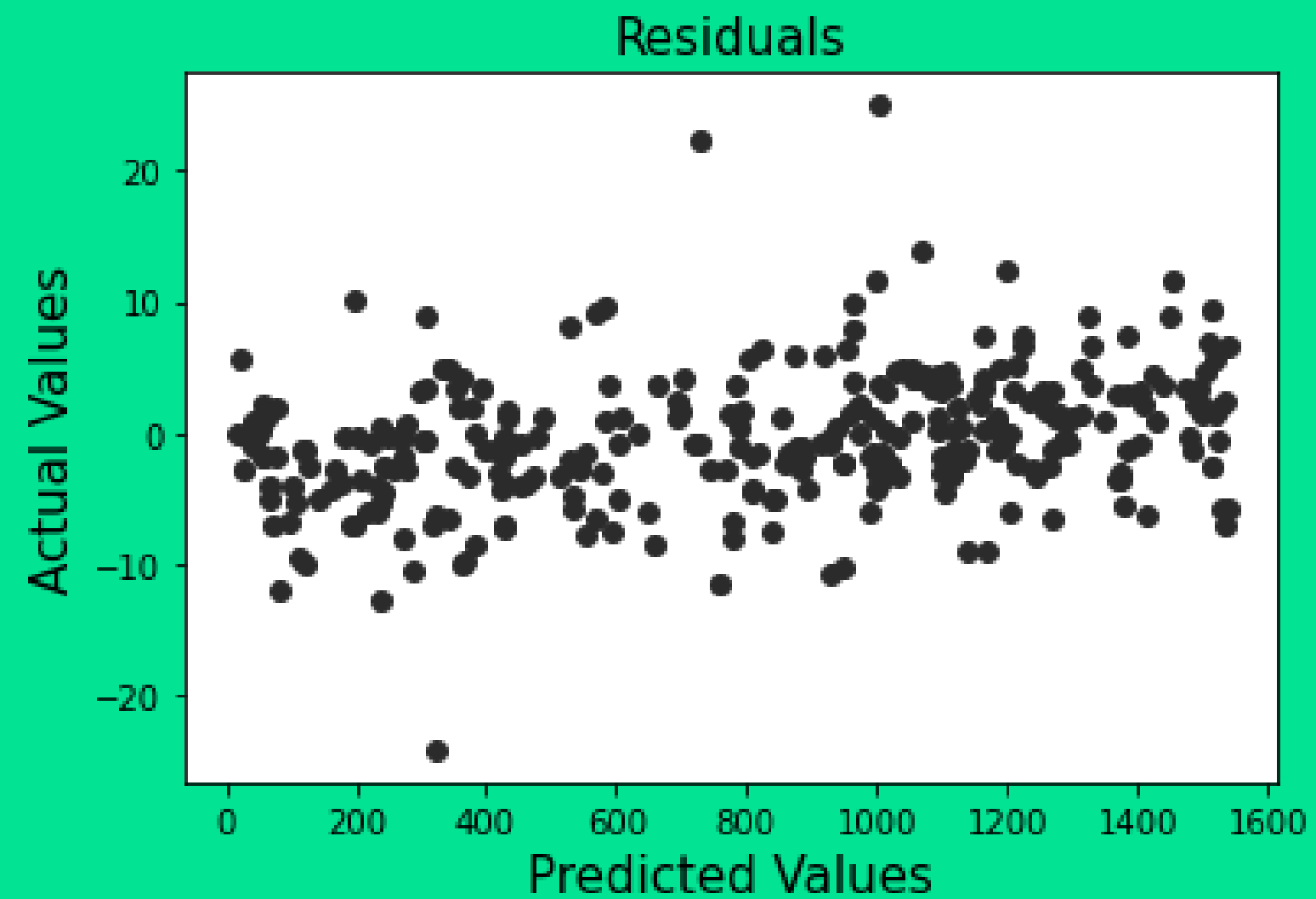
R2 Train Score: 0.972

RMSE Train Score: 2.49

R2 Test Score: 0.868

RMSE Test Score: 5.32

LASSO  
LASSO  
LASSO  
LASSO



## RANDOM FOREST – AFTER LR/LASSO

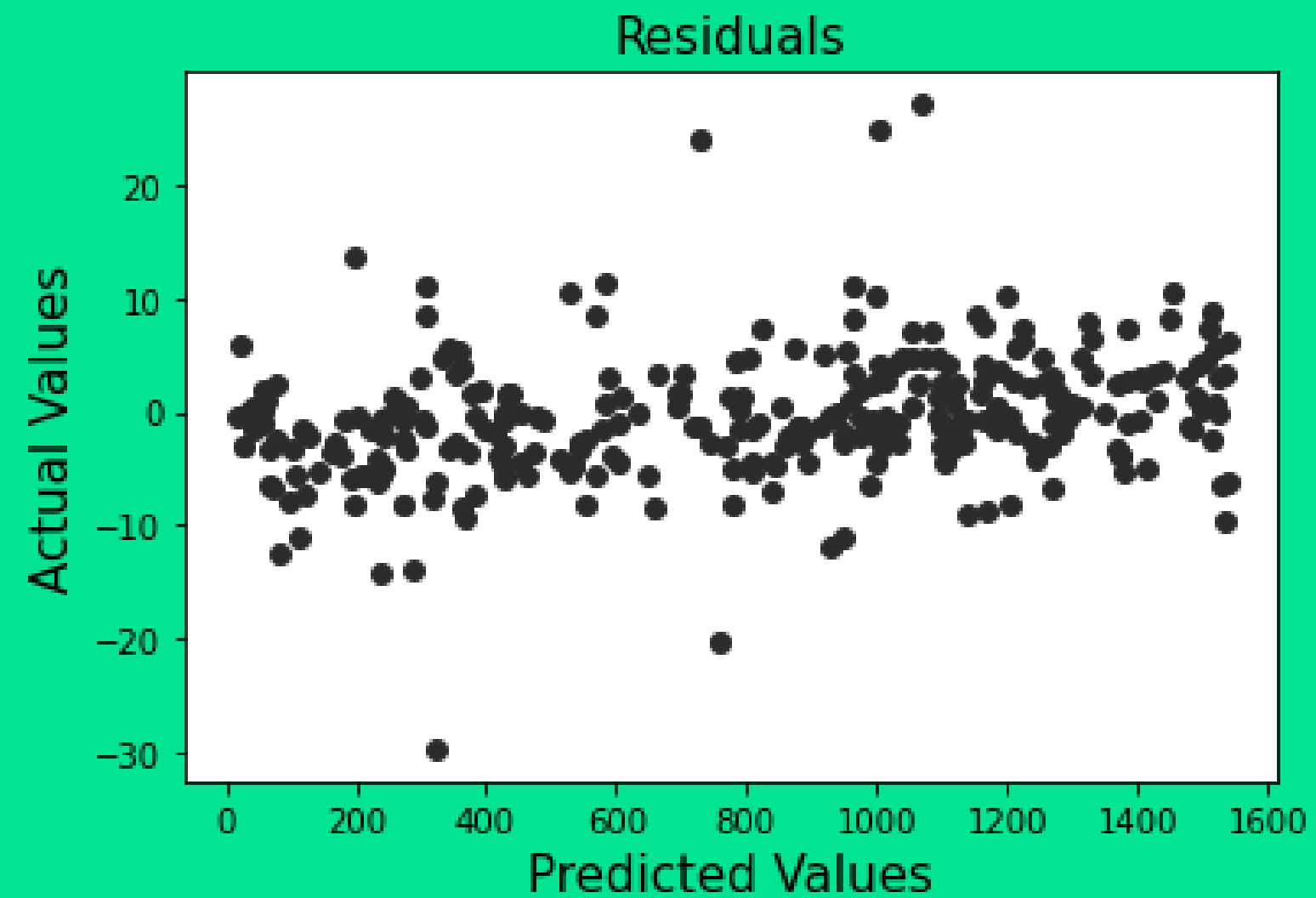
R2 Train Score: 0.972

RMSE Train Score: 2.48

R2 Test Score: 0.872

RMSE Test Score: 5.25

random forest gridsearch  
random forest gridsearch  
random forest gridsearch  
random forest gridsearch



## RANDOM FOREST – WITH GRIDSEARCH

R2 Train Score: 0.903

RMSE Train Score: 4.65

R2 Test Score: 0.848

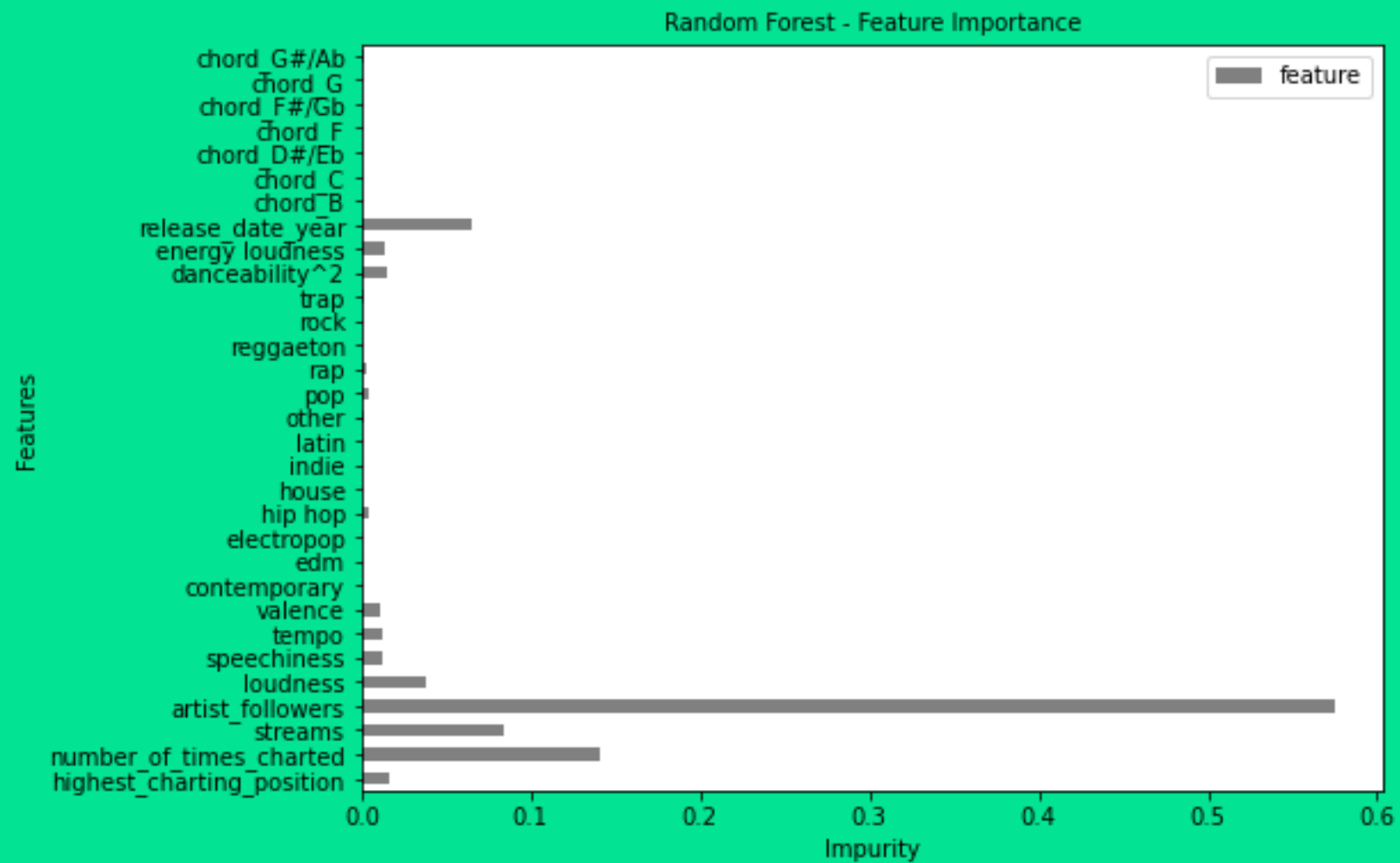
RMSE Test Score: 5.71

# Feature Importance

...

## RANDOM FOREST

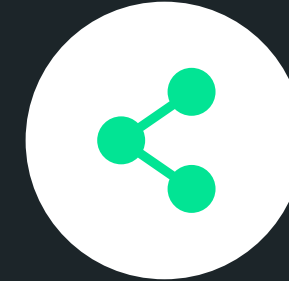
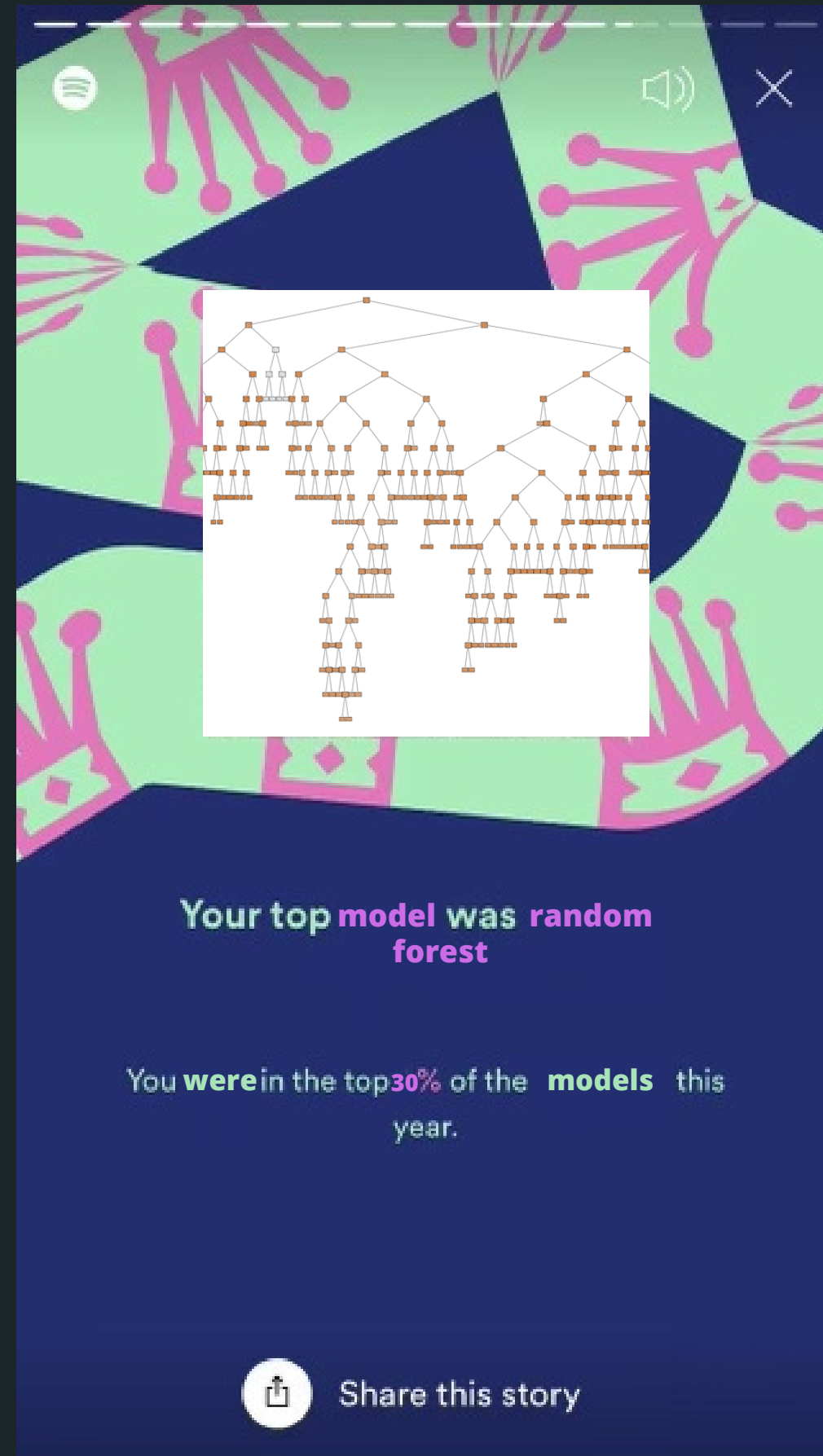
- artist followers
- number of times charted
- streams
- release date (year)



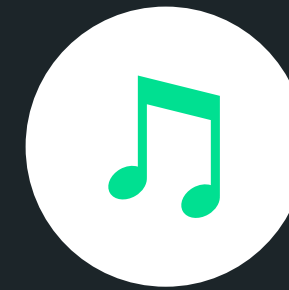
# Conclusion



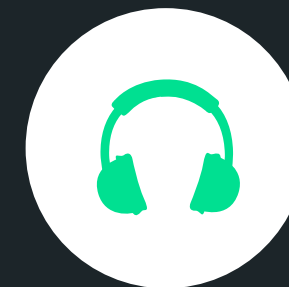
You trained 10 different models, but things got pretty serious with one...



Improvement of ~14% in R2 score and ~30% in RMSE



Feature Importance: artist followers, streams, release date



Recommendation: focus on artist following, release date, but also specific audio features



# Sources

- [https://www.investopedia.com/terms/r/r-squared.asp#:~:text=R%2Dsquared%20\(R2\),variables%20in%20a%20regression%20model.](https://www.investopedia.com/terms/r/r-squared.asp#:~:text=R%2Dsquared%20(R2),variables%20in%20a%20regression%20model.)
- <https://static.javatpoint.com/tutorial/machine-learning/images/linear-regression-in-machine-learning.png>
- <https://www.mastersindatascience.org/wp-content/uploads/tree-graphic.jpg>
- [https://upload.wikimedia.org/wikipedia/commons/7/76/Random\\_forest\\_diagram\\_complete.png](https://upload.wikimedia.org/wikipedia/commons/7/76/Random_forest_diagram_complete.png)
- <https://imgflip.com/meme/290748647/Spotify-Wrapped>
- [https://www.reddit.com/r/outerwilds/comments/r6hydf/spotify\\_wrapped\\_2021\\_guess\\_who\\_my\\_top\\_artist\\_was/](https://www.reddit.com/r/outerwilds/comments/r6hydf/spotify_wrapped_2021_guess_who_my_top_artist_was/)