



VINUNIVERSITY

Analyzing Taylor Swift's Discography

Data Visualizaton Project 1 Report

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INTRODUCTION

- Motivation: analyze a dataset about Taylor Swift's songs to gain insights into her success strategies.
- Importance: music production, marketing, and cultural impact, benefiting aspiring artists and music industry professionals.
- Dataset: Taylor Swift's studio albums, EPs, singles, and re-releases,
 - Song features, Album release dates, Metacritic scores, User scores.
 - Originates from the Genius lyrics database and Spotify's API

QUESTION 1:

How do features such as `danceability`, `energy`, `valence`, and `tempo` correlate with Taylor Swift's albums' critical and public reception?

INTRODUCTION

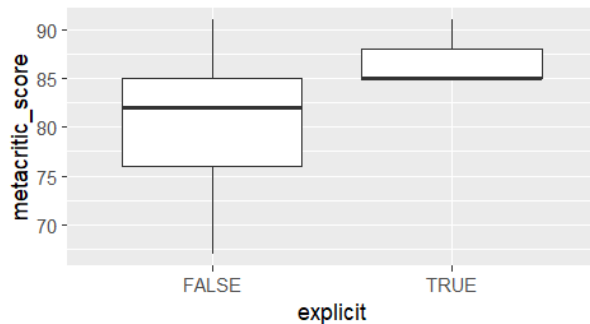
- By analyzing features such as lyrics' explicitness, tempo and key, we hope to uncover patterns in Taylor Swift's songs that contribute to her music's popularity.
- The overall dataset is as follows:
 - Three columns represent the index
 - 11 numerical features
 - 6 categorical features
 - 2 target numerical variables

APPROACH

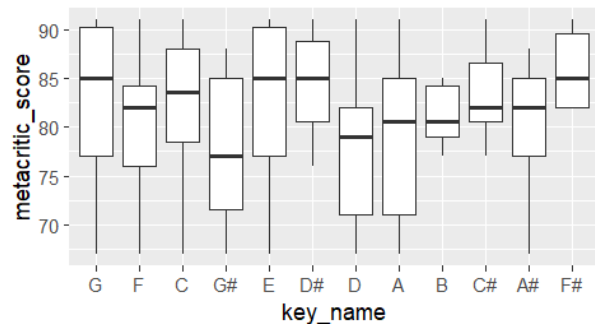
- The features incorporated in our analysis are split into two types: numerical and categorical features.
- Numerical variables: correlation matrix.
 - Common way to reveal relationships between numerical or binary variables
- Categorical variables: faceted box plots.
 - Using a correlation matrix is not recommended for non-binary categorical variables.
 - We also use bar charts to visualize class distribution, since class imbalance can impact our observations.

CATEGORICAL VARIABLES

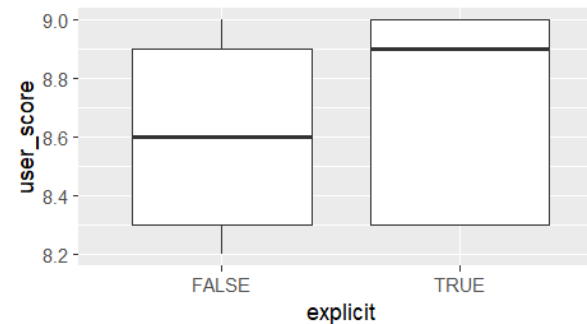
Distribution of metacritic_score by explicit



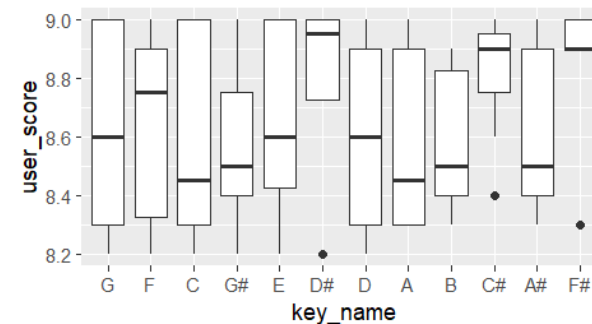
Distribution of metacritic_score by key_name



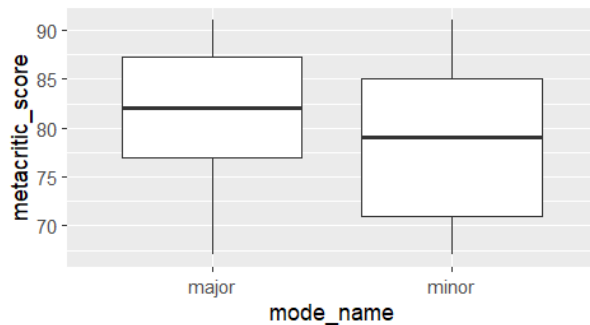
Distribution of user_score by explicit



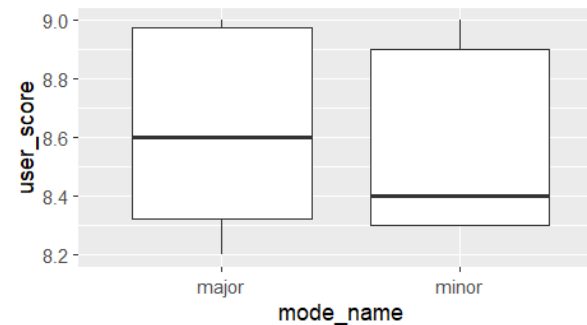
Distribution of user_score by key_name



Distribution of metacritic_score by mode_name

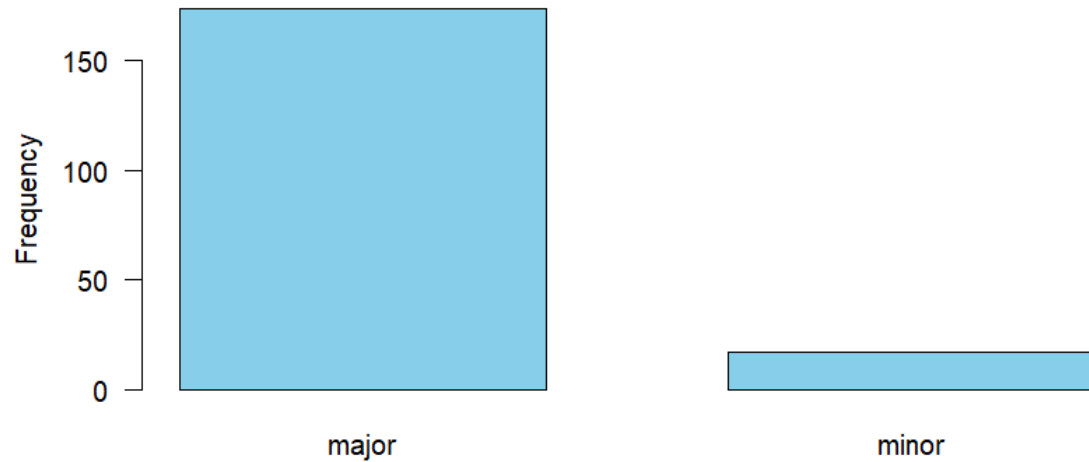


Distribution of user_score by mode_name

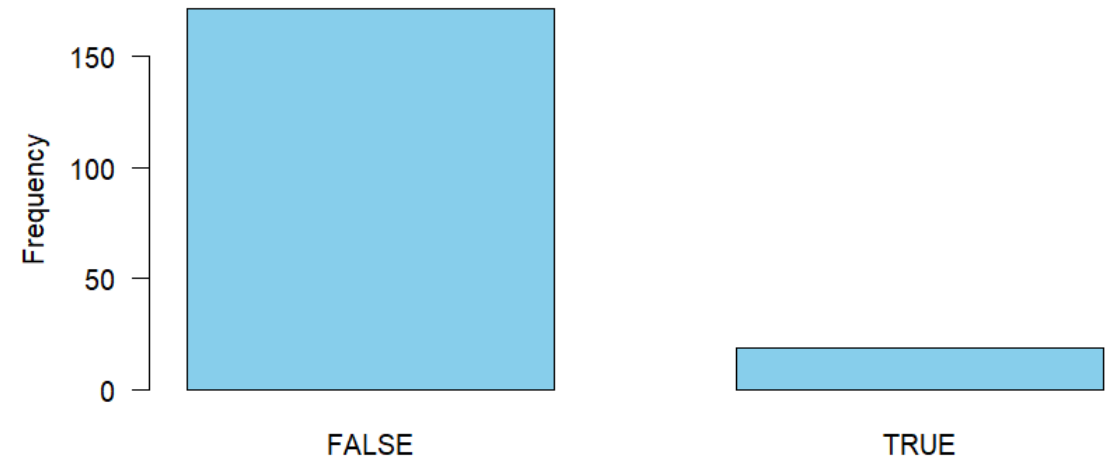


CATEGORICAL VARIABLES

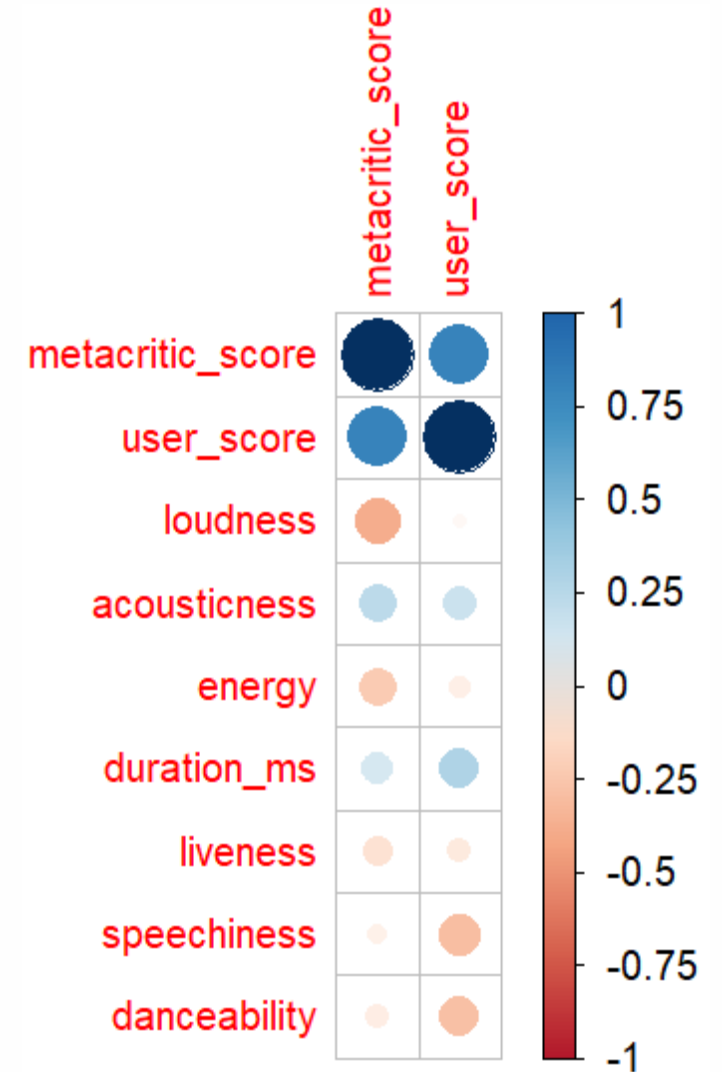
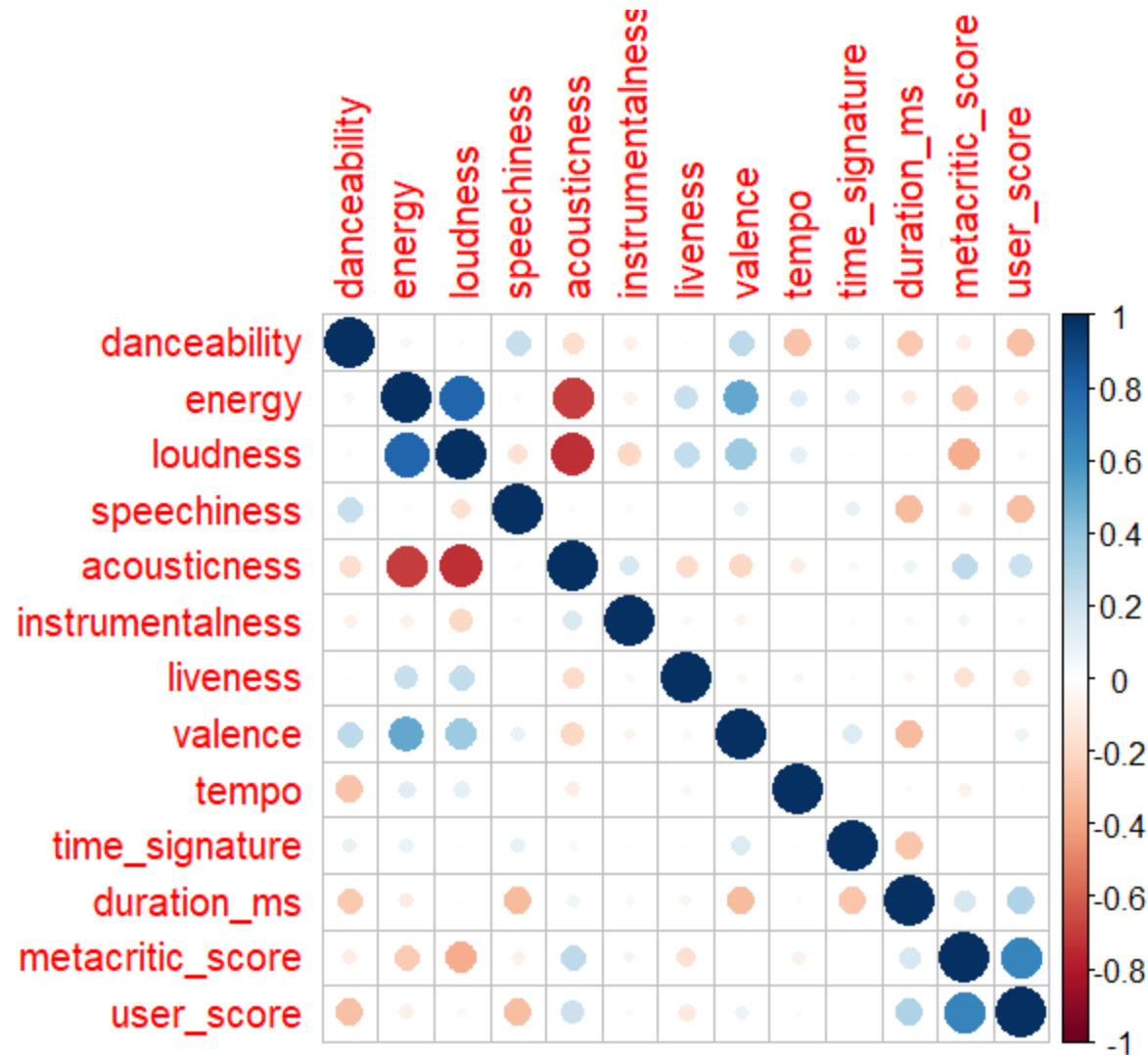
Frequency of mode_name



Frequency of explicit



NUMERICAL VARIABLES



SUMMARY

- Explicit and upbeat songs tend to score higher
 - Cannot be certain due to data imbalance
- Songs in C#, D#, and F# tend to get the highest scores with little variance, while those in keys such as A, C, and G often score lower and have more variance.
- Loudness, energy, liveness, speechiness, and danceability are frowned upon.
- Longer and more acoustic songs are favoured.





QUESTION 2

**HOW HAS TAYLOR SWIFT'S MUSICAL
STYLE EVOLVED OVER HER CAREER?**

INTRODUCTION

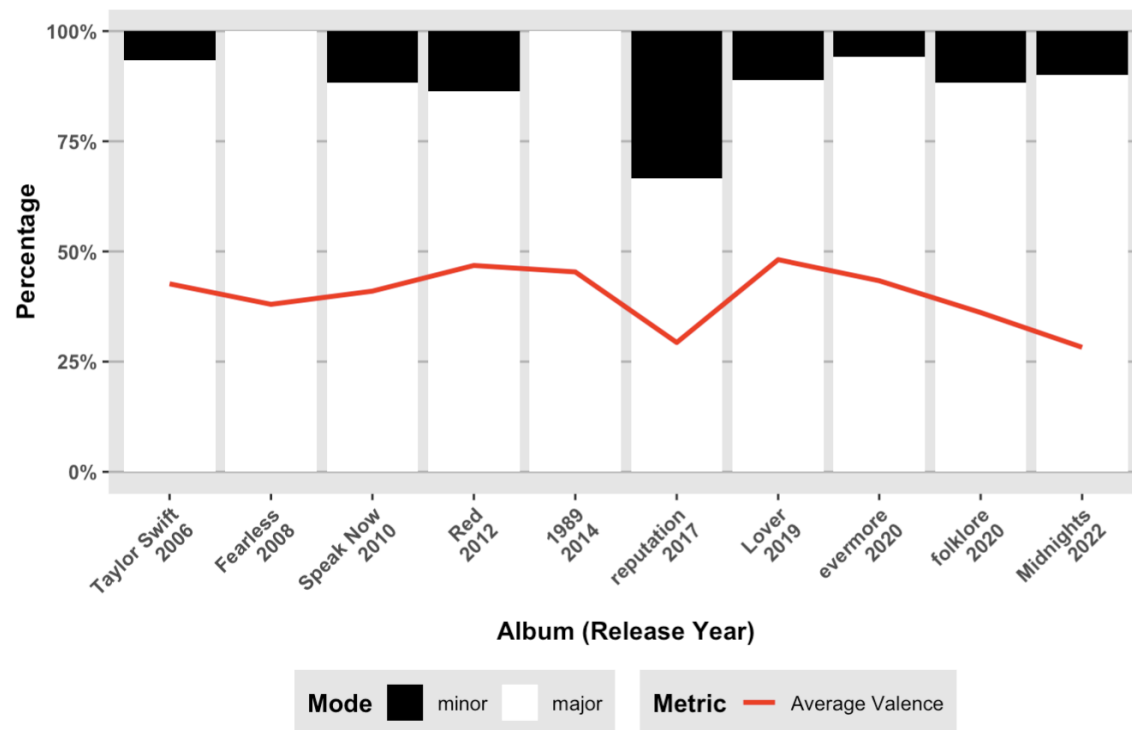
- The question seeks to investigate the changes of Taylor Swift's musical style throughout her career.
- By examining and comparing audio features across her albums, we can identify potential trends and changes in her music.

APPROACH

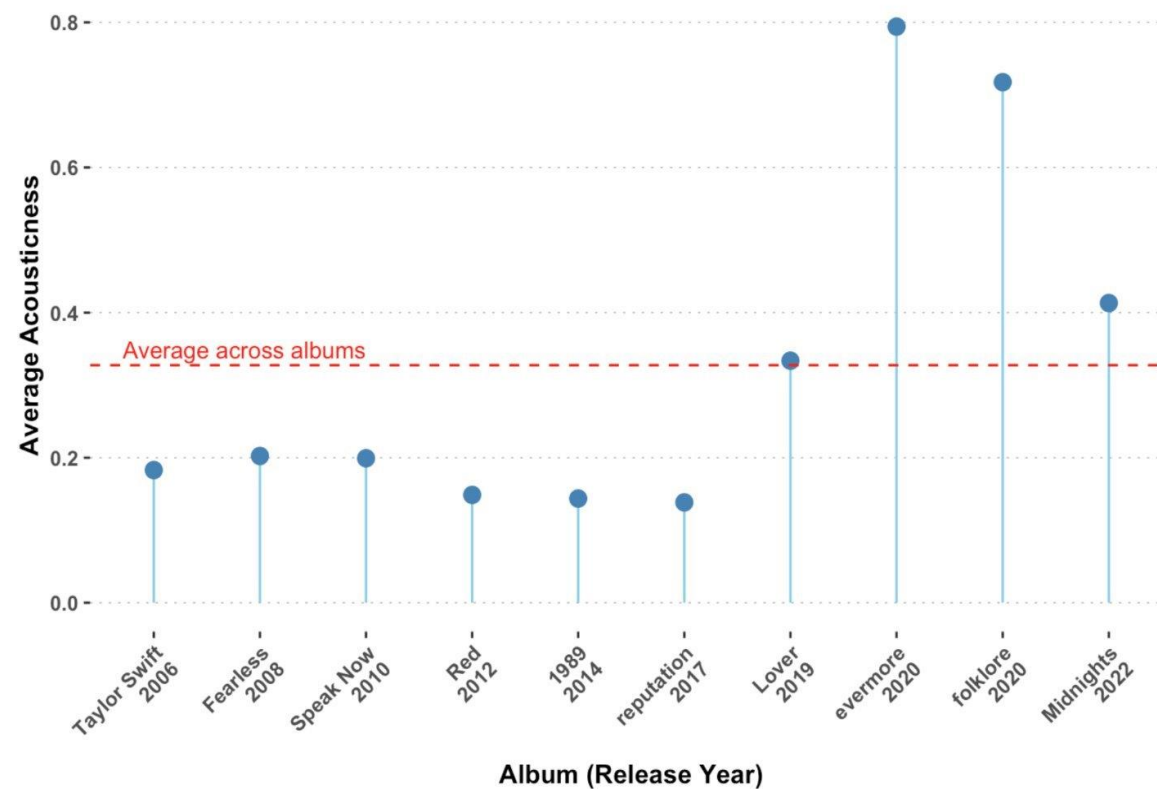
- Our approach includes the use of three types of graph:
 - (1) a relative frequency bar plot:
 - Compare the distribution of songs in minor versus major modes for each album.
 - (2) line chart:
 - Show average valence.
 - (3) a lollipop plot:
 - Compare average acousticness scores across albums.

Mean Valence and Relative Frequency of Modes for Each Album

How sad or happy is Taylor Swift music over time?



Average Acousticness of Each Album



DISCUSSION

- **Major Modes Dominance:** major modes for a brighter sound.
- **Low Valence Scores:** valence scores are under 50%, indicating moderate positivity.
- **"Reputation" as an Outlier:** 2017's "Reputation" has more **minor mode** tracks, resulting in the **lowest valence score** and making it her saddest album.
- **Increase in Acousticness Post-"Reputation":** A significant rise in acousticness in albums after "Reputation," with "Evermore" showcasing a fourfold increase.
- **Influence of External Events:** Changes in style potentially linked to the 2016 Kanye West scandal, reflecting a shift towards more authentic sounds.