

# Analyzing Spotify Trends with R

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## Spotify Dataset



- ▶ The Spotify Tracks Dataset contains audio features and popularity information for songs across many genres
- ▶ The dataset was obtained from Kaggle

# Introduction

- ▶ Dataset: “Spotify Tracks Dataset” from Kaggle
- ▶ Contains audio features and popularity data for Spotify songs
- ▶ Includes over 100000 tracks across various genres
- ▶ This project investigates whether certain song characteristics are associated with higher Spotify popularity

## Loading the Data

```
##           track_name track_genre popularity danceability
## 1           Comedy    acoustic         73         0.676
## 2   Ghost - Acoustic  acoustic         55         0.420
## 3           To Begin  acoustic         57         0.438
## 4 Can't Help Falling  acoustic         71         0.266
## 5                   Hold On acoustic         82         0.618
## 6   Days I Will Remember acoustic         58         0.688
```

- ▶ `read.csv()` imports the dataset into R
- ▶ `head()` displays the first few rows of the dataset

## Summary Statistics

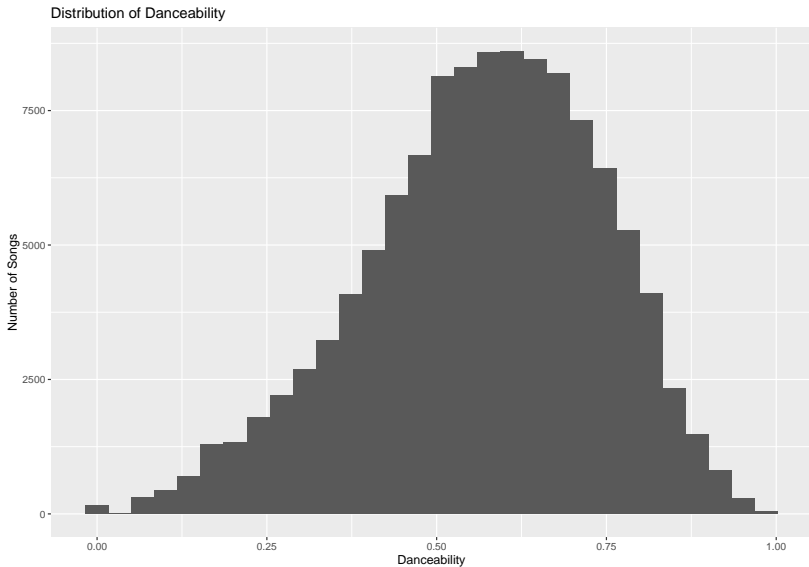
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00   17.00   35.00   33.24  50.00  100.00
```

```
## [1] 33.23854
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.0000  0.4560  0.5800  0.5668  0.6950  0.9850
```

- ▶ `summary()` gives descriptive statistics like minimum, median, mean, and maximum
- ▶ `mean()` calculates the average popularity score
- ▶ Danceability measures how suitable a song is for dancing
- ▶ The average popularity score is about 33, which suggests many songs in the dataset are not extremely mainstream
- ▶ The median danceability is around 0.58, meaning many songs are moderately danceable
- ▶ Danceability values range widely, showing substantial variation across songs

# Distribution of Danceability



## Distribution of Danceability

- ▶ This histogram shows the distribution of danceability scores across Spotify songs
- ▶ Most songs fall between about 0.4 and 0.8 danceability, suggesting moderately danceable songs are most common
- ▶ Very low and very high danceability scores appear less frequently

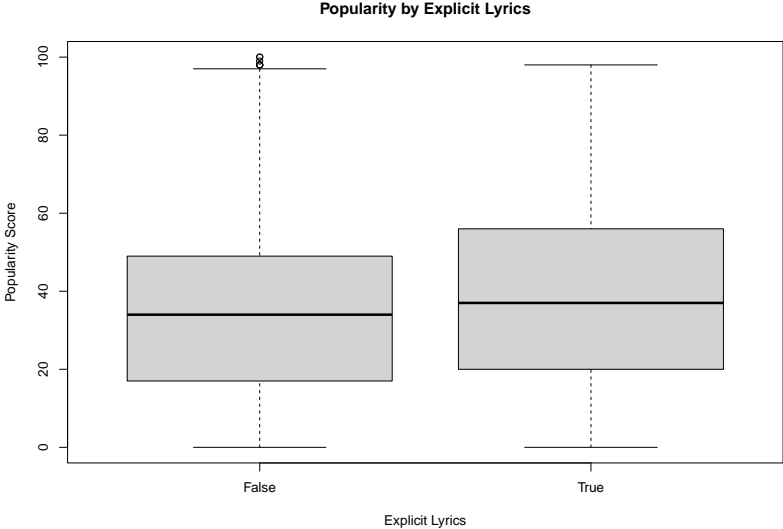
## Explicit vs Non-Explicit Songs

```
##  
##  False    True   <NA>  
## 104253   9747     0
```

```
##  
## False   True  <NA>  
##  0.91  0.09  0.00
```

- ▶ About 91% of songs in the dataset are not explicit
- ▶ Only around 9% are labeled explicit
- ▶ This suggests non-explicit songs make up the majority of Spotify tracks in this dataset

# Popularity by Explicit Lyrics



## Popularity by Explicit Lyrics

- ▶ This boxplot compares popularity for explicit and non-explicit songs
- ▶ The median popularity for explicit songs appears slightly higher
- ▶ Explicit songs also show a somewhat wider spread of popularity scores
- ▶ However, the distributions overlap substantially, suggesting explicit content alone does not determine popularity

## Average Popularity by Genre

##	track_genre	popularity
## 82	pop-film	59.283
## 66	k-pop	56.896
## 16	chill	53.651
## 95	sad	52.379
## 45	grunge	49.594
## 56	indian	49.539
## 6	anime	48.772
## 34	emo	48.128
## 98	sertanejo	47.866
## 81	pop	47.576

- ▶ The genres with the highest average popularity include pop-film, k-pop, chill, and sad
- ▶ K-pop appears near the top of the rankings, suggesting strong listener engagement on Spotify
- ▶ The results suggest popularity on Spotify may vary substantially across musical genres

# Conclusion

- ▶ The project demonstrated how data visualization and summary statistics can help reveal patterns in large datasets
- ▶ Overall, the analysis suggests that song characteristics such as danceability, genre, and explicit content may be related to popularity on Spotify
- ▶ Although no single factor fully determines popularity, the analysis shows how R can be used to explore trends and relationships within real-world data