

### What Makes a Playlist Successful?

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# Agenda

- Project Goal
- Strategy
- Defining Success Metrics
- Feature Candidates for Successful Playlists
- Data Exploration and Visualization
- Regression Analysis
- Conclusions
- Next Steps



### Project Goal (Listen to Playlist More and Longer)

Business Goal: Provide the most successful playlists to Spotify customers to

- usage <sub>main goal</sub>
- customer satisfaction indirect result
- revenue indirect result

Technical Goals: Quantify what makes playlist successful

- How to measure the successfulness of the playlist?
- What are the characteristics of successful playlists?



# Strategy

Quick Look at Stream Data

Defining Success Metrics (Dependent Variable)

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Data Preparation

Data Preparation

Data Exploration and Visualization Analysis

Linear Regression

**Conclusions** 



# Looking at Steam Data

	playlist_uri	owner	streams	stream30s	 genre_3	mood_1	mood_2	mood_3
0	spotify:user:36069af6af076ccd9e597184a67b68c9:	36069af6af076ccd9e597184a67b68c9	27	27	 Country & Folk	Peaceful	Romantic	Somber
1	spotify:user:d1144a65b1c31c5f9f56b94f831124d5:	d1144a65b1c31c5f9f56b94f831124d5	0	0	 Alternative	Excited	Yearning	Defiant
2	<pre>spotify:user:6b7fbed9edd6418ddd3b555bba441536:</pre>	6b7fbed9edd6418ddd3b555bba441536	4	2	 -	Lively	Upbeat	Romantic
3	spotify:user:580b98725077a94c3c8d01d07390426b:	580b98725077a94c3c8d01d07390426b	12	12	 Pop	Excited	Aggressive	Defiant
4	spotify:user:1305d39070c95d161cc502e15014897d:	1305d39070c95d161cc502e15014897d	20	4	 Electronica	Excited	Defiant	Yearning
	***				 			
403361	spotify:user:4672952d42bdd93b9215ce9a40394ea6:	4672952d42bdd93b9215ce9a40394ea6	18	6	 Dance & House	Excited	Yearning	Energizing
403362	spotify:user:28c4378e099b4843f5dd42bb848c78ea:	28c4378e099b4843f5dd42bb848c78ea	0	0	 Rap	Sensual	Excited	Brooding
403363	spotify:user:1c54302dc7e610a10c51eed81e26a168:	1c54302dc7e610a10c51eed81e26a168	0	0	 Alternative	Brooding	Defiant	Sophisticated
403364	<pre>spotify:user:adc973443cdf1abecdfb4244e530d451:</pre>	adc973443cdf1abecdfb4244e530d451	0	0	 Latin	Defiant	Energizing	Aggressive

[403366 rows x 25 columns]



### Looking at Steam Data

#### **Descriptive Statistics for Some Features**

Number of Playlists*	402967
Number of Owners	314898
AVG Number of Tracks of Playlists	202
AVG Number of Albums of Playlists	88
AVG Number of Artists of Playlists	84
AVG Number of Tokens of Playlists	2
AVG Number Streams of Playlists (Today)	21
AVG Number Streams30s per Playlists (Today)	12 (57%)
AVG Number Streams30s per Playlists (This Month)	354

AVG DAU per Playlist	1
MIN DAU per Playlist	0
MAX DAU per Playlist	18290
AVG WAU per Playlist	5
MIN WAU per Playlist	0
MAX WAU per Playlist	71891
AVG MAU per Playlist	15
MIN MAU per Playlist	2
MAX MAU per Playlist	206756



<sup>\*</sup> Excluded 399 Spotify Playlists

### **Defining Success Metrics**

#### When can we say that the playlist is successful?

- Users listen to the successful playlists a lot
- Users listen to the successful playlist for a longer time (longer than 30 seconds)

#### What are possible levels for the success?

- Success at playlist level
- Success at the playlist owner level



### **Defining Success Metrics**

#### **Short Term**

DAUs

Number of daily active users with streams more than 30 seconds from playlist

#### WAUs

Number of weekly active users with streams more than 30 seconds from playlist

#### Streams30s

Number of streams from playlist today with more than 30s

#### **Long Term**

MAUs

Number of monthly active users with a stream over 30 seconds from playlist this month. Or total number of active users from the playlist in the last 2 months (mau\_both\_months)

Monthly\_streams30s

Number of streams over 30 seconds by playlist

Monthly\_owner\_streams30s

Number of streams over 30 seconds by playlist owner this month

Today: WAU

In Future: Combination of DAU, WAU, MAU, Streams30s, Monthly\_streams30s, Monthly\_streams30s

### **Candidate Features**

### **Playlist Variety**

- Number of Tracks
- Number of Artists
- Number of Albums

#### Playlist Owner Descriptors

- Number of playlists per owner (proxy for owner effort)
- Owner has more than 1 playlist (proxy for owners activity)

#### Playlist Categorical Features

- Moods: Leading Mood
- Genres: Leading Genres
- Tokens Sentiment (Senitment Analysis for tokens)

#### **Playlist Ratios**

- Streams30s to Streams Ratio
- Streams30 to Monthly Streams Ratio
- Monthly\_owner\_stream30s to monthly\_stream30sRatio

### Hypothesis

### Varity Matters

Genre, moods, number of tracks, albums, artists

#### **Effort Matters**

Length of title

#### **Sentiment Matters**

Sentiment Analysis moods, genre, tokens

### **Engagement Matters**

Number of Playlists Owned by User



### Data Preparation

#### **Checking for Outliers**

- Removing 399 Playlists of Spotify with above than average stream
- Removing the rows from 99th percentile

#### **Checking for Missing Values**

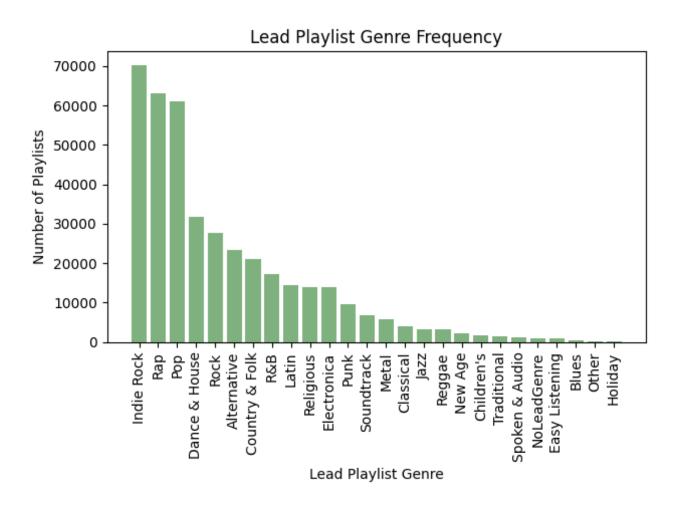
There are no null values in the data but there are missing values, especially in Mood, Genre variables

#### **Information Retrieval and Sentiment Analysis for Tokens**

- Converting text to list of strings
- Sentiment Score low: Negative vs high: Positive (NLP on tokens)



#### **Playlists Leading Genre**



#### **Popular Genres Leading Playlists**

 3 most popular leading genres across all playlists: "Indie Rock", "Rap", "Pop"



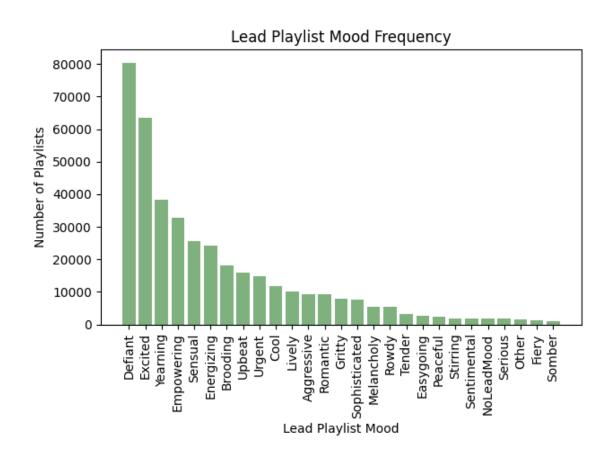
Playlists from these 3 genres are more successful?



 Binary features from these 3 for genres for future predictive model



#### **Playlists Leading Mood**



#### **Popular Moods Leading Playlists**

2 most popular leading moods across all playlists: "Defiant", "Excited"

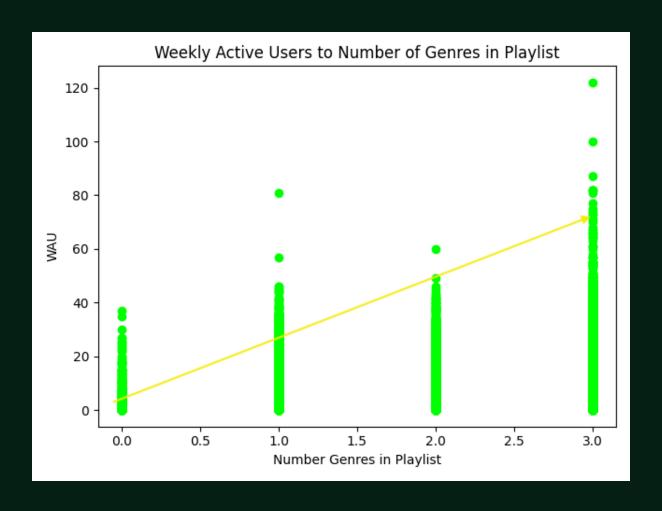


Playlists with these 2 moods are more successful



 Binary features from these 2 moods for future predictive model

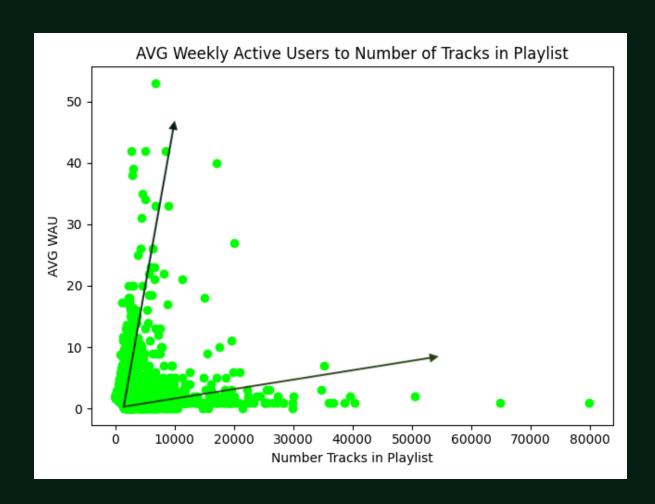




#### Variety in genre matters

- Positive relationship between number of genres and WAU
- Playlists with 3 genres have higher WAU than playlists with less than 3 genres

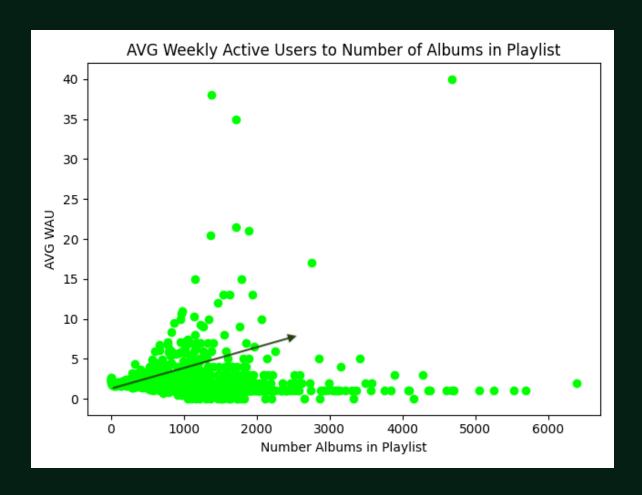




Variety in tracks doesn't matter

 No clear relationship between number of tracks in playlists and its WAU

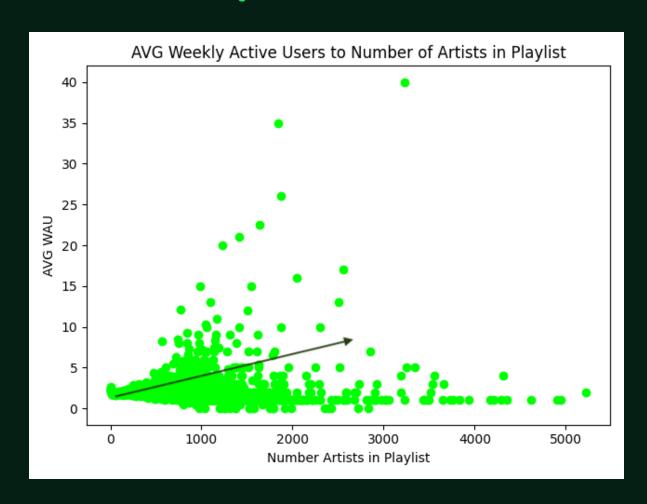




#### Variety in Albums matters

- Positive relationship between number of albums in playlist and WAU
- Playlists with more albums show higher
   WAU than playlists with very small
   amount of albums
- But this effect wears off but correlation is not causation





#### Variety in Artists matters

- Positive relationship between number of artists in playlist and WAU
- Playlists with more artists show higher
   WAU than playlists with very small
   amount of artists
- But this effect wears off but correlation is not causation



#### **Sentiment Analysis using NLP**

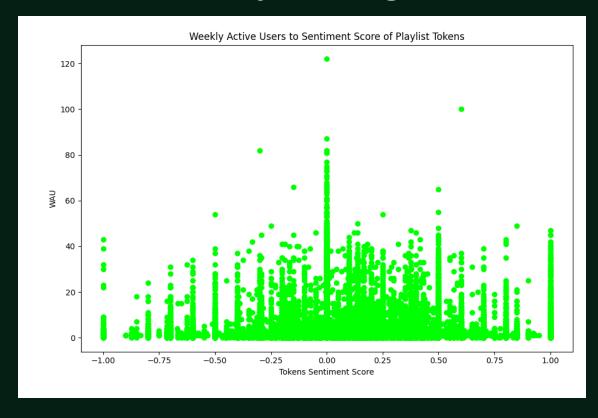
Most Positive Tokens with Highest Sentiment Score (1)

Most Negative Tokens with Lowest Sentiment Score (-1)

	tokens	sentiment_score_tokens		tokens	sentiment_score_tokens
26832	["best"]	1.0	170780	["awful", "playlist"]	-1.0
332300	["best", "rock", "roll"]	1.0	49591	["awful", "feelings"]	-1.0
41139	["awesome"]	1.0	112677	["spank", "rock", "everything", "boring", "eve	-1.0
121742	["awesome"]	1.0	65024	["make", "nasty"]	-1.0
21059	["greatest", "hits"]	1.0	216320	["make", "work", "terrible", "mix"]	-1.0
	•••			***	
146580	["best"]	1.0	73398	["bleak"]	-1.0
258191	["awesome", "sauce"]	1.0	212903	["worst", "enemy"]	-1.0
104147	["kalapana", "best"]	1.0	323691	["songs", "terrible"]	-1.0
398253	["anita", "baker", "best", "anita", "baker"]	1.0	357532	["insane", "clown", "posse"]	-1.0
5463	["awesome", "mix"]	1.0	126980	["terrible", "pop"]	-1.0



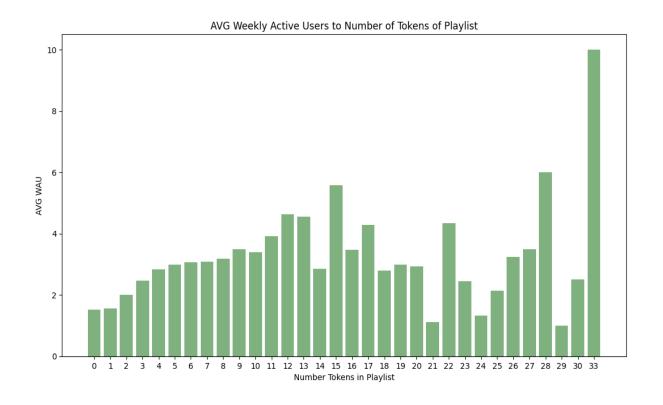
#### **Sentiment Analysis using NLP**



#### Sentiment of Playlist Tokens doesn't matters

 No clear relationship between sentiment of the playlists tokens and its WAU





#### **Effort Doesn't really matters**

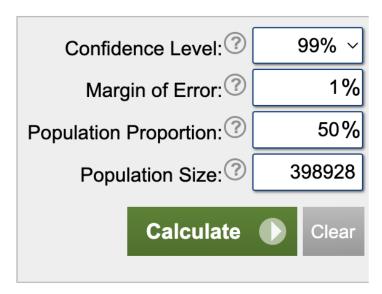
- No clear pattern convincing that the length of the tokens (effort proxy) has an impact on the usage of the playlist
- There is some positive pattern from 0-11 as num tokens but after this effect wears off



### Linear Regression (preliminary)

Sample size: 15975

This means 15975 or more measurements/surveys are needed to have a confidence level of 99% that the real value is within ±1% of the measured/surveyed value.





#### OLS Regression Results

OLS Regression Results								
Dep. Variable:	wau	R-squ	ared (uncente	0.852				
Model:	0LS	OLS Adj.R-squared (un						
Method:	Least Squares	east Squares F-statistic:				6546.		
Date:	Tue, 22 Mar 2022	Prob	(F-statistic)	:	: 0.			
Time:	19:45:40	Log-L	ikelihood:		-28607.			
No. Observations:	15975	AIC:			5.724e+04			
Df Residuals:	15961	BIC:	BIC:			5.735e+04		
Df Model:	14							
Covariance Type:	nonrobust							
=====================================	===========	======	=======================================			=======================================		
	coef	std err	t	P> t	[0.025	0.975]		
genre_1_Indie Rock	0.0212	0.033	0.648	0.517	-0.043	0.085		
genre_1_Rap	-0.0089	0.041	-0.217	0.828	-0.089	0.071		
genre_1_Pop	-0.1739	0.034	-5.119	0.000	-0.240	-0.107		
mood_1_Defiant	-0.0483	0.037	-1.319	0.187	-0.120	0.023		
mood_1_Excited	0.0081	0.033	0.248	0.804	-0.056	0.072		
stream30s	0.0172	0.001	27.587	0.000	0.016	0.018		
n_tracks	-2.015e-05	2.5e-05	-0.805	0.421	-6.92e-05	2.89e-05		
n_local_tracks	-4.616e-05	0.000	-0.294	0.768	-0.000	0.000		
n_artists	-0.0002	0.000	-0.744	0.457	-0.001	0.000		
n_albums	0.0007	0.000	2.789	0.005	0.000	0.001		
num_genres	0.2295	0.009	26.395	0.000	0.212	0.247		
num_tokens	0.0041	0.009	0.470	0.638	-0.013	0.021		
sentiment_score_token	s -0.0127	0.059	-0.215	0.830	-0.128	0.103		
users	0.1099	0.000	228.984	0.000	0.109	0.111		
Omnibus:	8386.991	Durbi	n-Watson:					
Prob(Omnibus):	0.000	Jarqu	ue-Bera (JB):	14094583.336				
Skew:	-0.927	-	_	0.00				
Kurtosis:	148.504	Cond.	No.	3.30e+03				

### Linear Regression

- **Pop Genre:** significant and negative effect
- Streams30s: significant and positive effect
- Num\_albums: significant and positive effect
- Num\_genres: significant and positive effect
- Users: significant and positive effect



### Conclusions

### What makes playlist successful?

Variety in albums

Variety in genres



### Next steps

Statistical Significance

More Advanced Regression

Combined and Customer Success Metric

More Features

Supervised Machine
Learning (Ensemble or NN
Models for Predictions)
Improving Playlist Search
or Recommender
Algorithm
(Reranker Model &
Impression Discounting)

A/B Testing to compare new vs old Playlist Search Engine or Recommender System





# Thank you!