# "What we talk about when we talk about successful playlist"

-- Hong Wu (Interview Assignment)

#### **DISCLAIMERS**

With limitation/restriction from the given dataset, some estimation/projection are based open data source which may differ from actual Spotify cases. All charts, visualizations and conclusions are pure for demonstrating personal data analysis capabilities.

#### **Assumptions on Data**

With provided data dictionary, all data is on daily basis, projection made on time goes by 24 hours, 7 days a week , 31 days per month, simply for calculation convenience, open data is also used.

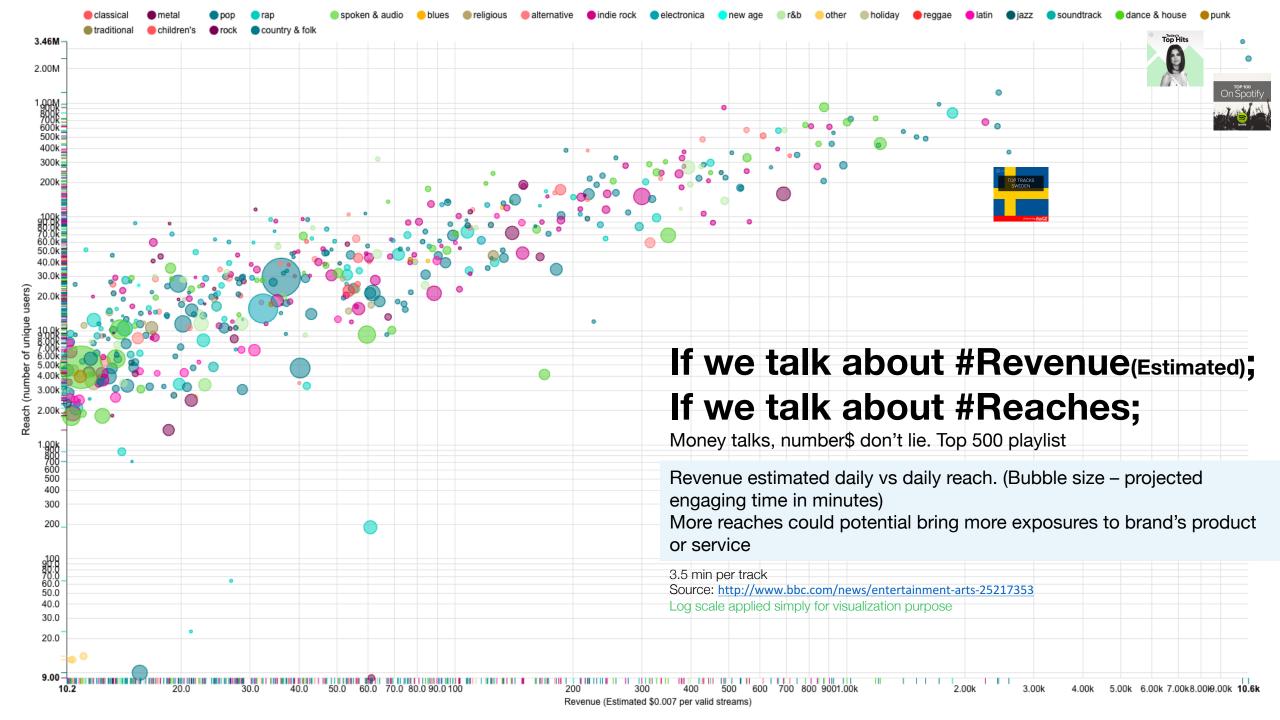
All scripts, code, and data is available in the following repo https://github.com/thehongwu/playlist\_summary\_analysis

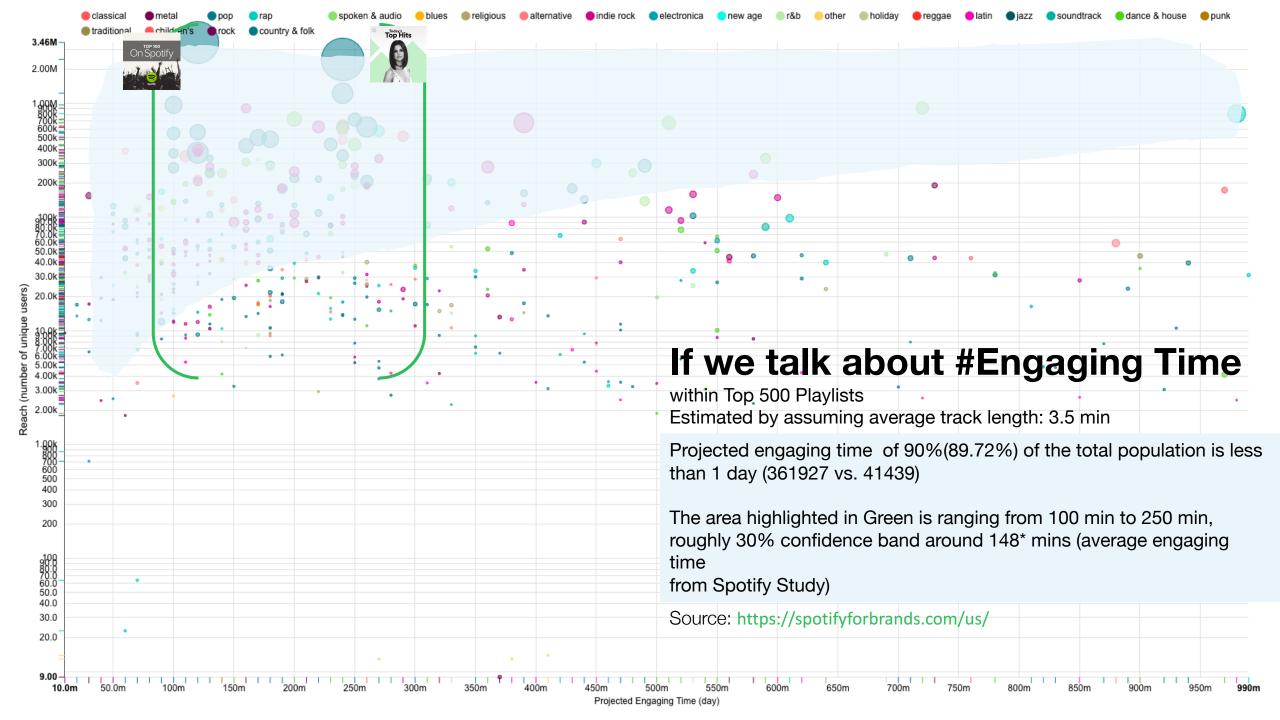
Will delete after interview

### **Technical Sheet - Audience: Data/Tech Team**

A little bit on Data Processing

- mysql
- Spotify API: https://developer.spotify.com/web-api/console/get-playlist/#complete
- Python, superset
- NLTK





new age

spoken & audio

religious

classical

reggae

dance & house

**country & folk** 



blues

If we talk about #Genre

Word cloud generated from top 1000 playlist

Pop, Indie rock, Dance & House, and Rap are the top tier guys and the rest

rap

holiday

latin

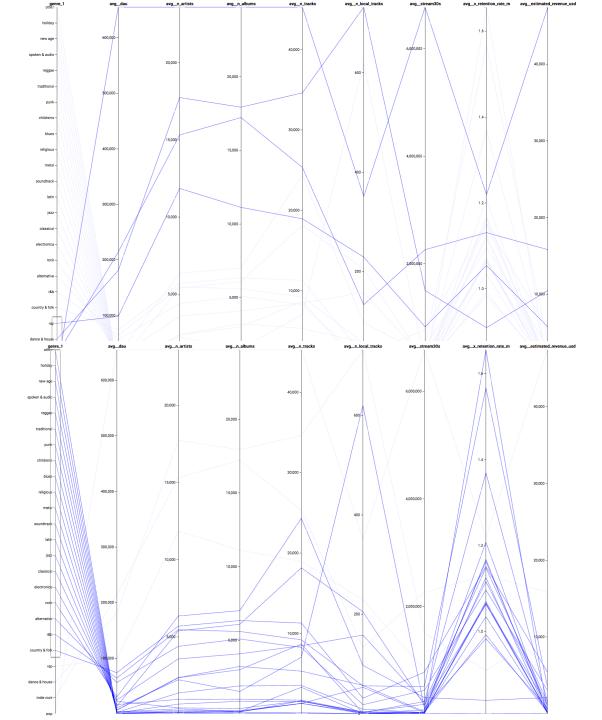
electronica





easy listening





### If we talk about #Genre

Parallel coordinates generated from top 1000 playlist Which verifies the result from word cloud:

"More Active user, more valid stream, more engaging time"

DAU is a "good splitor" which indictors "Potential Two Clusters" in terms of common "metrics":

- {Pop, Indie Rock, Dance & House, Rap}
- Vs. {The rest}

tender

peaceful lively

sensual

empowering



defiant yearning

romantic

fiery

<sub>ırgent</sub> energizing

If we talk about #Mood

Word cloud generated from top 1000 playlist

other

Top moods:

- Excited
- defiant
- yearning
- melancholy

melancholy

gritty

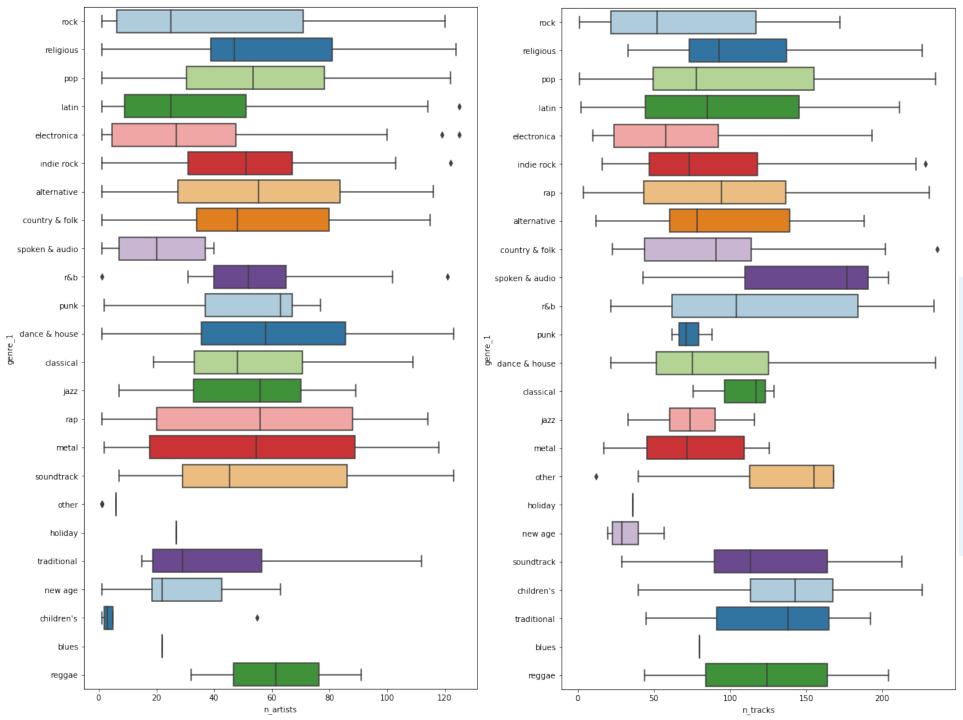
aggressive

serious

sophisticated

rowdy

cool



## If we talk about #Tracks/Artists

Boxplot generate from top 1000 playlist with extreme/anomalies data taken out

We can generalize what a playlist looks like from this perspective, some highlights:

**pop**: 81 tracks - 60 artist similar

number of album

indie rock: 78 tracks - 58 artist

similar number of album

dance & house: 83 tracks - 65 artist

similar number of album

rap: 95 tracks - 75 artist similar

number of album

# A formula of a "successful" playlist leading to high engaging and revenue-generation.

#### NOTE:

Purely from this datasets, and data/metrics perspective along with assumptions



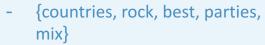


One more thing...

If we talk about #MUSIC
If we talk about #USERS
If we talk about #INTEREST

#### If we talk about #Tastes

With tokens available, we are able to extract top 5 taste from top 1000 playlists



- {pop, hit, music, summer, folk}
- {music, Christian, new, parties, update}
- {playlist, ultimate, workout, best, summer}
  - {song, love, classic, new, countries}





Majority of the active users could be characterized as "Explorers": loves country music, party and its summer!!! which possibly opens new opportunities for Brands and advertiser to do targeted marketing, trying new stuffs, new restaraunts, things related to party, travel, concert etc.



"What we "REALLY" talk about when we talk about successful playlist"

**Affinity Between Content and Audience**