Music Recommendation System

Group: Double Eleven

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1. EDA

We have 2 datasets: one of users' playlists, one of songs' features.

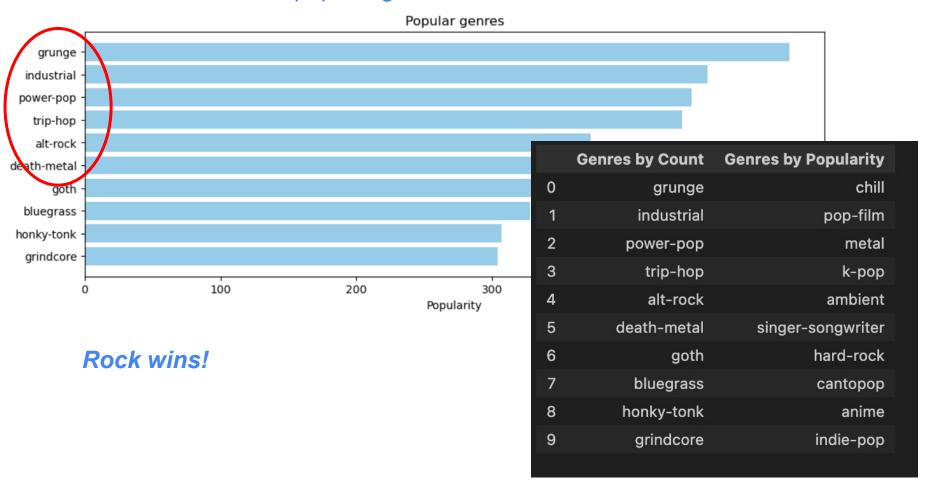
		usei	_id		artist	is				tr	ack_nam	е	playlist_name
0	9cc0cfd4d7d788510	2480dd99e7a90	0d6	E	Elvis Costell	lo	(The A	ngels Wann	a Wear	Му)	Red Shoe	s HA	RD ROCK 2010
1	9cc0cfd4d7d788510	2480dd99e7a90	0d6 Elvis C	ostello & Th	e Attraction	ns (What's	So Funn	y 'Bout) Pea	ce, Love	e An	d Unders.	НА	RD ROCK 2010
2	9cc0cfd4d7d788510	Dd6	Tiffany Page		je	7 Years Too Late				е НА	RD ROCK 2010		
3	9cc0cfd4d7d788510	2480dd99e7a90	0d6 Elvis C	ostello & Th	e Attraction	ıs			Acciden	its V	Will Happe	n HA	RD ROCK 2010
4	9cc0cfd4d7d788510	0d6	Elvis Costello		lo	Alison			n HA	RD ROCK 2010			
	track_id	artists	album_name	track_name	popularity	duration_ms	explicit	danceability	energy		loudness	mode	speechiness
5	SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	73	230666	False	0.676	0.4610		-6.746	0	0.1430
	4qPNDBW1i3p13qLCt0Ki3A	Ben Woodward	Ghost (Acoustic)	Ghost - Acoustic	55	149610	False	0.420	0.1660		-17.235	1	0.0763
	1iJBSr7s7jYXzM8EGcbK5b	Ingrid Michaelson;ZAYN	To Begin Again	To Begin Again	57	210826	False	0.438	0.3590		-9.734	1	0.0557
	6lfxq3CG4xtTiEg7opyCyx	Kina Grannis	Crazy Rich Asians (Original Motion Picture Sou	Can't Help Falling In Love	71	201933	False	0.266	0.0596		-18.515	1	0.0363
	5vjLSffimiIP26QG5WcN2K	Chord Overstreet	Hold On	Hold On	82	198853	False	0.618	0.4430		-9.681	1	0.0526

Duplicate songs exist in either dataset:

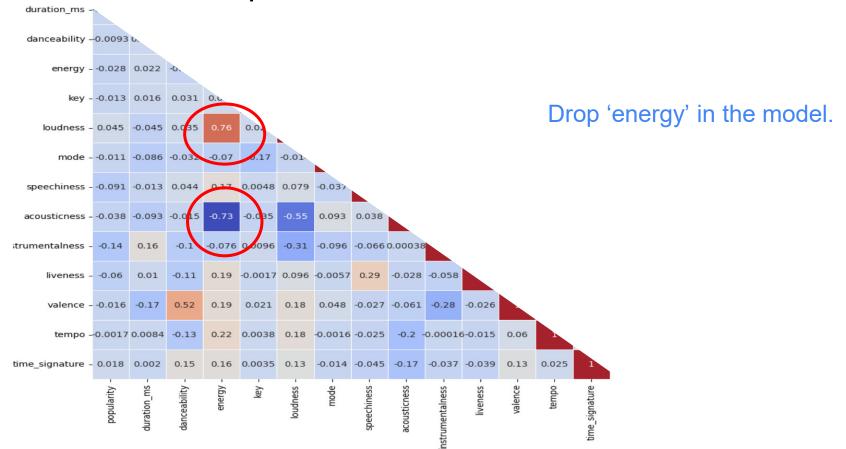
track_id	artists	album_name	track_name	popularity	duration_ms	explicit	danc
5SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	73	230666	False	
5SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	73	230666	False	
5SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	73	230666	False	
5SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	73	230666	False	

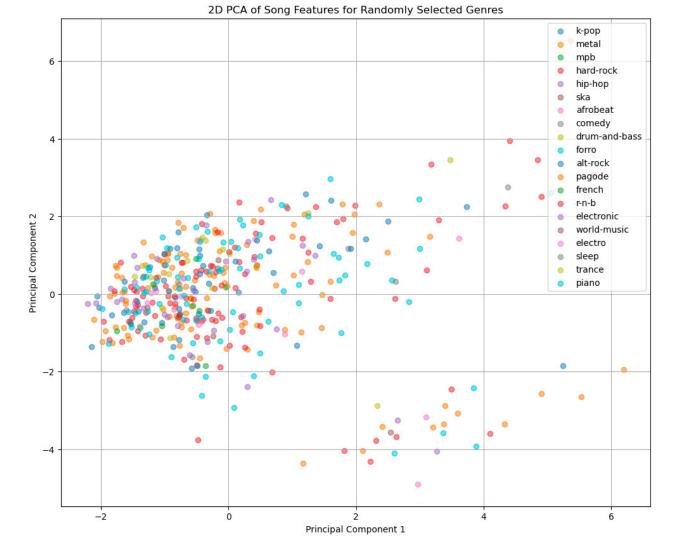
Drop songs with same artists and name for the model.

First let's take a look at popular genres.



Heatmap of numerical features





2. Model - Main idea

- Use vectors to represent songs
- For a user
 - Get his/her taste vector based on the song he/she likes
 - Identify k song vectors closest to the taste vector
 - Recommend them!

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Use vectors to represent songs For a user ased on the song he/she likes Get hi st to the taste vector Ider How to define the vector for each song?



	Numerical features	Text features
song1		
song2		

Text Vectors	Numerical Vectors

Our target

The songs liked by the same user should be closer!

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Words in similar contexts



Word vectors with closer distance

Sentence 1	[,
Sentence 2		,
Sentence n		,

Song features Word2vec liked by same user

Feature vectors with closer distance

User 1	,
User 2	,
User n	1

Example

```
interest = 'taylorswift'
   similar words = vectorizer.model.wv.most similar(interest, topn=10)
   for word in similar_words:
       print(word)
    0.0s
('justinbieber', 0.8739178776741028)
('jasonderulo', 0.7950416207313538)
('selenagomezthescene', 0.7633613348007202)
('vanessacarlton', 0.7504104971885681)
('christinaperri', 0.7465274333953857)
('krisallen', 0.7457529902458191)
('lifehouse', 0.7431434988975525)
('shaniatwain', 0.7302798628807068)
('neyo', 0.7079509496688843)
('gleecast', 0.7035359144210815)
```

Hit@ K

A user's playlist with N songs

