CSE343: ML Project Monsoon 2021

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OK Computer: Employing Machine Learning to analyse the impact of different parameters in foretelling the next earworm





Motivation

It is **formulaic** — and that's not a bad thing. Music is a social phenomenon that's influenced and modeled off prior art. As a result, popular music genres often do become homogenized, recycling the same themes, structure and lyrics over and over again. 02-Jun-2016

https://www.washingtonpost.com > in-theory > 2016/06/02

Opinion | Your favorite songs all sound the same - The ...

https://livinglifefearless.co > features > breaking-formul...

Breaking Down the Formula of 'Formulaic' Music - LIVING LIFE ...

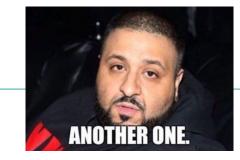
Predictability in **music** encompasses all aspects of a **song**, from chorus to coda. **Popular music** is, as a rule, predictable **music**. Think, Kesha's "Tik Tok" and ...

https://www.washingtonpost.com > 2014/06/27 > wann...

Wanna write a pop song? Here's a fool-proof equation - The ...

27-Jun-2014 — Does the **pop song formula** mean that all popular music sounds the same? Clearly not, and there will be plenty of songs performed at the ...

Motivation



- In the age of TikTok and Reels, it is claimed that the songs regularly reaching the top of the charts are getting repetitive, almost following a pattern that guarantees its success.
- Some artists (Another one, DJ KHALED!) have even gone on to make rather braggadocious claims that they know well before the release of their song whether it's going to be a hit or not.
- We wish to answer these burning questions What is it, after all, that seems to make a song more popular than others, and can we accurately predict the popularity of a track.

Literature Review



Hit Song Science Once Again a Science by Ni et al.

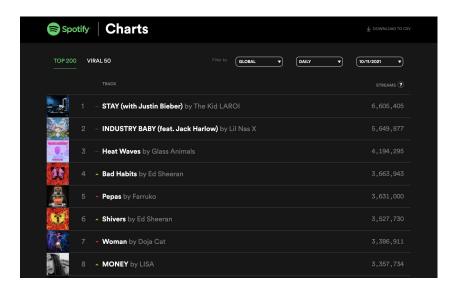
- This paper talks about using the shifting perceptron algorithm to classify top 5 songs from the rest from the weekly charts over the last 50 years.
- Features like tempo, time signature, song duration and loudness, coefficient of variance of loudness and harmonic simplicity were used.
- The EchoNest API has been used to extract the features for the 5947 unique songs that were collected from the Official Charts Company.
- Successfully investigates the change in musical taste over the years which separates it from its previous studies. Uses classifiers only.

Literature Review

- Predicting Music Popularity on Streaming Platforms by Araujo et al.
- This paper talks about the methodology to predict if a song will appear on Spotify's Top 50 Global ranking after a certain amount of time.
- Features such as entry's rank, duration, explicit flag and daily popularity score were used for the same.
- SVM, Random Forests and Gaussian Na ive-Bayes models were used to classify whether a song is successful or not based on its appearance on the Spotify Top 50 Global charts.

Dataset

We were after the latest possible data available for our tests. Thus we set out to create our own dataset and made use of Spotify Charts data from 2017-21, which we procured from the Spotify Charts site.



Dataset

The Spotify ID was then used with the Spotify Web API to garner the set of acoustic features that it provides. As a result, we were able to generate a dataset of 4247 songs, each with a total of 23 features -

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1 Id		Rank	Track	Artist	Streams	Week	Album_name	Explicit	Track_number	Artist_follow	Artist_genre	Acousticness	Danceability	Energy	Instrumental L	iveness	Loudness	Speechiness	Tempo	Mode	Key	Valer
2 http:	s://open		1 Starboy	The Weeknd	25734078	06/01/17	Starboy		1 1	31348348	['canadian co	0.165	0.681	0.594	3.49E-06	0.134	-7.028	0.282	186.054		1	7
3 http:	s://open		2 Closer	The Chainsm	23519705	06/01/17	Closer		0 1	17742887	['dance pop',	0.414	0.748	0.524	0	0.111	-5.599	0.0338	95.01		1	8
4 http:	s://open		3 Rockabye (fe	Clean Bandit	21216399	06/01/17	Rockabye (fe		1	4296325	['dance pop'	0.406	0.72	0.763	0	0.18	-4.068	0.0523	101.965		0	9
5 http:	s://open		4 Let Me Love	DJ Snake	19852704	06/01/17	Encore		13	7312319	['dance pop'	0.0784	0.476	0.718	1.02E-05	0.122	-5.309	0.0576	199.864		1	8
6 http	s://open		2 I Don,Äôt W	ZAYN	30752312	17/02/17	I Don,Äôt Wa		0 1	15423979	['dance pop',	0.0631	0.735	0.451	1.30E-05	0.325	-8.374	0.0585	117.973		1	0
7 http:	s://open		6 Don't Wanna	Maroon 5	18064374	06/01/17	Don't Wanna		0 1	30323494	['pop', 'pop r	0.338	0.783	0.623	0	0.0975	-6.126	0.08	100.048		1	7
8 http:	s://open		7 Fake Love	Drake	17037036	06/01/17	More Life		1 20	54405324	['canadian h	0.105	0.928	0.481	0	0.176	-9.35	0.287	134.007		0	9
9 http:	s://open		7 Say You Wor	James Arthu	18269129	13/01/17	Back from th		2	7893527	['pop', 'post-	0.695	0.358	0.557	0	0.0902	-7.398	0.059	85.043		1	10
10 http:	s://open		9 24K Magic	Bruno Mars	16736035	06/01/17	24K Magic		0 1	29942000	['dance pop',	0.034	0.818	0.803	0	0.153	-4.282	0.0797	106.97		1	1
11 http:	s://open		9 Feel It Com	The Weeknd	17465511	13/01/17	Starboy		18	31348348	['canadian co	0.426	0.773	0.819	0	0.0679	-5.946	0.118	92.99		0	0
12 http:	s://open		11 Black Beatle	Rae Sremmı	16130702	06/01/17	SremmLife 2		1 5	6359203	['hip hop', 'm	0.142	0.794	0.632	0	0.128	-6.163	0.0649	145.926		1	0
13 http:	s://open		12 One Dance	Drake	15958402	06/01/17	Views		12	54405324	['canadian h	i 0.00784	0.791	0.619	0.00423	0.351	-5.886	0.0532	103.989		1	1
14 http:	s://open	:	13 Chantaje (fe	Shakira	14458068	06/01/17	El Dorado		3	21467047	['colombian	0.187	0.852	0.773	3.05E-05	0.159	-2.921	0.0776	102.034		0	8
15 http:	s://open		14 Cold Water (Major Lazer	14278458	06/01/17	Cold Water (0 1	6184098	['dance pop',	0.0736	0.608	0.798	0	0.156	-5.092	0.0432	92.943		0	6
16 http:	s://open		13 Call On Me -	Starley	16094980	20/01/17	Call On Me (6	132367	['aussietroni	0.0604	0.67	0.838	0.000611	0.159	-4.031	0.0362	104.998		1	0
17 http:	s://open		16 In the Name	Martin Garri	13936848	06/01/17	The Martin 6		2	14595128	['dance pop'	0.0592	0.49	0.485	0	0.337	-6.237	0.0406	133.889		0	4
18 http:	s://open		9 Bad and Bou	Migos	18937768	03/02/17	Culture		1 4	11354618	['atl hip hop'	0.061	0.927	0.665	0	0.123	-5.313	0.244	127.076		1	11
40 1			10 CIJ- T- CIJ-	A-1 C	12002756	00/04/47	D			C43004F4	ri I	0.0400	0.040	0.720	^	0 202	F 003	0 247	150 145		^	•

Dataset

We extracted the lyrics using Google's Engine API. And then performed Topic Modelling using LDA to extract features from the Lyrics Dataset.



lyrics.py (base) Anshs-MacBook-Air:data extraction ansharora\$ python lyrics.py {'title': 'The Weeknd — Starboy Lyrics', 'lyrics': "[Verse 1]\nI'm tryna put you in the worst mood, ah\nP1 cleaner than your church shoes, ah\nMilli point two just to hurt you , ah\nAll red Lamb' just to tease you, ah\nNone of these toys on lease too, ah\nMade your whole year in a week too, yah\nMain bitch out of your league too, ah\nSide bitch out of your league too, ah\n\n[Pre-Chorus]\nHouse so empty, need a centerpiece\nTwenty racks a table, cut from ebony\nCut that ivory into skinny pieces\nThen she clean it with her face, man, I love my baby, ah\nYou talkin' money, need a hearing aid\nYou talkin' 'bout me, I don't see the shade\nSwitch up my style, I take any lane\nI switch up my cup, I kill any pain\n\n[Chorus]\nLook what you've done\nI'm a motherfuckin' starboy\nLook what you've done\nI'm a motherfuckin' starboy\n\n[Verse 2]\nEvery day, a nigga try to test me, ah\nEvery day, a nigga try to end me, ah\nPull off in that Roadster SV, ah\nPockets overweight, gettin' hefty, ah\nComin' for the king, that's a far cry, I\nI come alive: n the fall time, I\nThe competition, I don't really listen\nI'm in the blue Mulsanne, bumpin' New Edition\n\n[Pre-Chorus]\nHouse so empty, need a centerpiece\nTwenty racks a able, cut from ebony\nCut that ivory into skinny pieces\nThen she clean it with her face, man, I love my baby, ah\nYou talkin' money, need a hearing aid\nYou talkin' 'bout me, I don't see the shade\nSwitch up my style, I take any lane\nI switch up my cup, I kill any pain\n\n[Chorus]\nLook what you've done\nI'm a motherfuckin' starboy\nLook what you 've done\nI'm a motherfuckin' starboy\n\n[Verse 3]\nLet a nigga brag Pitt\nLegend of the fall, took the year like a bandit\nBought Mama a crib and a brand new wagon\nNow she h it the grocery shop lookin' lavish\nStar Trek roof in that Wraith of Khan\nGirls get loose when they hear this song\nA hundred on the dash get me close to God\nWe don't pray f or love, we just pray for cars\n\n[Pre-Chorus]\nHouse so empty, need a centerpiece\nTwenty racks a table, cut from ebony\nCut that ivory into skinny pieces\nThen she clean it with her face, man, I love my baby, ah\nYou talkin' money, need a hearing aid\nYou talkin' 'bout me, I don't see the shade\nSwitch up my style, I take any lane\nI switch up my cup, I kill any pain\n\n[Chorus]\nLook what you've done\nI'm a motherfuckin' starboy\nLook what you've done\nI'm a motherf uckin' starboy\nLook what you've done\nI'm a motherfuckin' starboy"}

Methodology



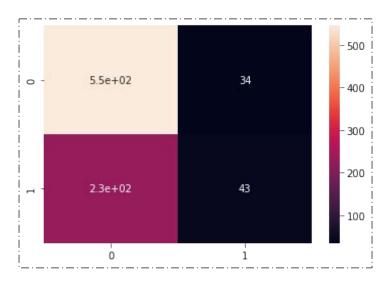
- We created labels for our use cases For Classification Problems, we used Hit
 (1) and Not-Hit (0) and for Regression Problems, we used a gradient scale from 0 to 1, 1 for songs that have charted at the top, 0.3 for songs that peaked at #200 and 0 for songs that have not charted yet.
- We carried out Classification Algorithms Logistic Regression and Support Vector Machines and have made use of Regression Algorithms - Logistic Regression, Regularized Regression and Support Vector Regression. We utilized Cross Validation techniques such as K-Fold and Holdout, and made use of GridSearchCV as well in cases to get the best hyperparameters.

Results and Analysis

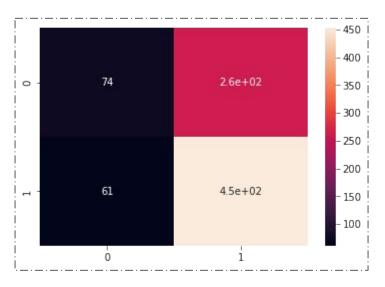


From the various models that we applied on our dataset, we found almost negligible correlation between any 2 features. Also, our logistic regression performs fairly well with accuracy peaking at 71% and precision at 84%. On apply LDA, we could extract 18 features with the best coherence score out of the 27 different range of features we tested on. Linear regression gave an average RMSE of **0.039**. Regularized Lasso Regression returned slightly better values here but not as good as Ridge Regression. Also, on doing the t-test and f-test on Linear regression model, we confirmed that our assumption of residual errors being normally distributed with mean 0 stands strong as the p-value calculated is above 5%.

Results and Analysis

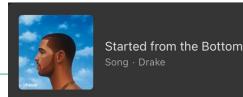


Keeping a threshold of 50, and kernel as linear.



Keeping a threshold of 100, and kernel as poly.

Timeline



So far we have been able to follow the proposed timeline exactly on a weekly basis (Till present week i.e. Week 6). As we continue to learn new processes in our course, we aim to make use of as many new skills as we can to arrive at the best result possible. We propose that we follow the path that our initial findings have shown in the coming weeks, other than using models that we had planned to use in the coming weeks.

Weeks 7-9 \rightarrow Implement remaining models on the existing and Million Song dataset, utilizing lyrics extracted features as well.

Weeks 10-11 → Analysis + Hyperparameter Tuning to give accurate results with less bias and variance

Week 12 → Report Writing

Contributions



Ansh	Lyrics Curation, Exploratory Data Analysis, Support Vector Regression, Report + Powerpoint
Jishnu	Data Curation, Linear Regression, Logistic Regression, Support Vector Classification, Report + Powerpoint
Nandika	Data Curation, Linear Regression, Support Vector Classification, Logistic Regression, Report + Powerpoint
Tushar	Lyrics Curation, Exploratory Data Analysis, Latent Dirichlet Allocation, Report + Powerpoint

