

# $X^i$ Records

Can Popularity Be Predicted?

# Introduction

- With Spotify being so well used is it possible to predict the popularity of song?
- Are there certain characteristics of a song that make it more popular?
- Can we predict whether a new artist has released a song to rival that of Ed Sheeran or Taylor Swift?

Let's See What the Data Says!

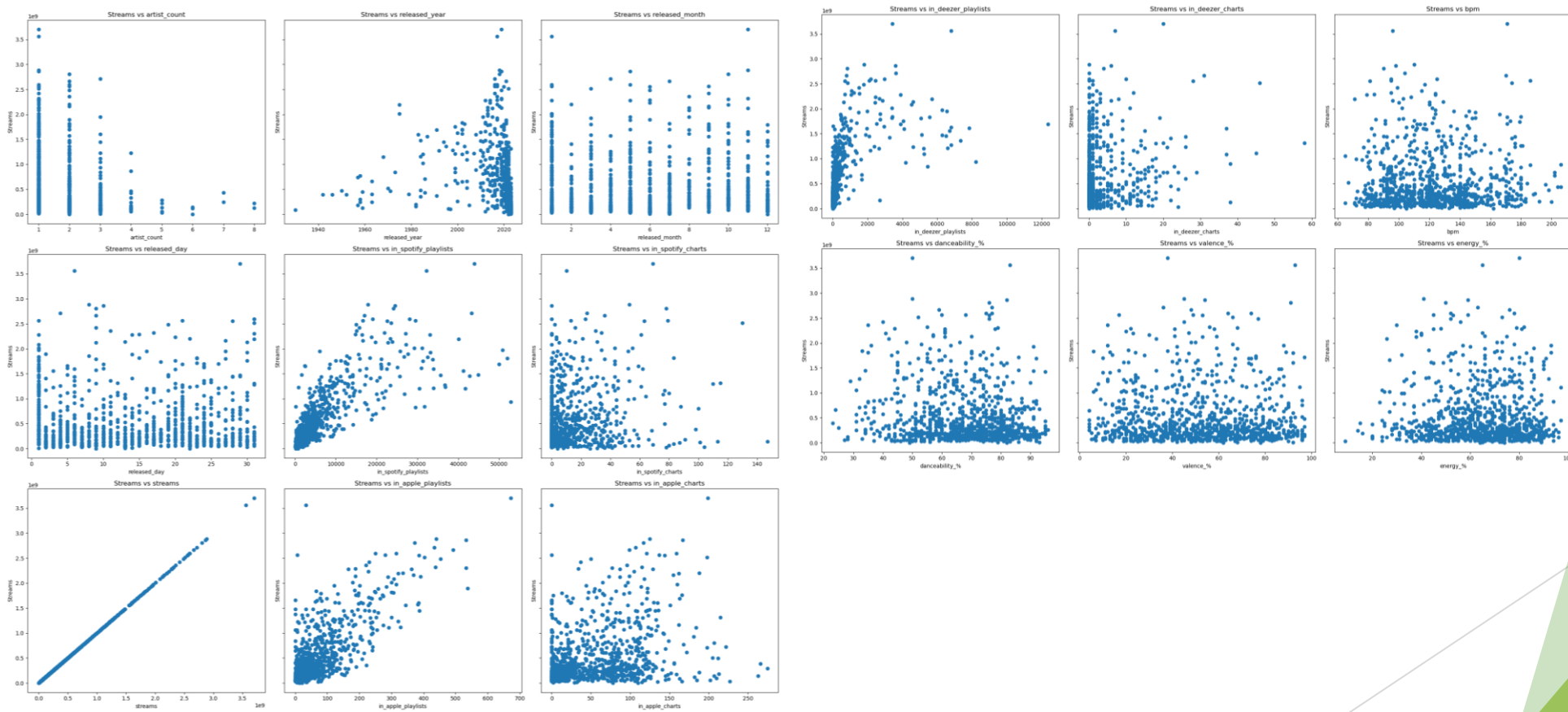
# The Data

The dataset I have sourced from Kaggle and is entitled “Most Streamed Spotify Songs 2023”. It has 943 rows and 24 columns with the titles:

- track\_name:** Name of the song
- artist(s)\_name:** Name of the artist(s) of the song
- artist\_count:** Number of artists contributing to the song
- released\_year:** Year when the song was released
- released\_month:** Month when the song was released
- released\_day:** Day of the month when the song was released
- in\_spotify\_playlists:** Number of Spotify playlists the song is included in
- in\_spotify\_charts:** Presence and rank of the song on Spotify charts
- streams:** Total number of streams on Spotify
- in\_apple\_playlists:** Number of Apple Music playlists the song is included in
- in\_apple\_charts:** Presence and rank of the song on Apple Music charts
- in\_deezer\_playlists:** Number of Deezer playlists the song is included in
- in\_deezer\_charts:** Presence and rank of the song on Deezer charts
- in\_shazam\_charts:** Presence and rank of the song on Shazam charts
- bpm:** Beats per minute, a measure of song tempo
- key:** Key of the song
- mode:** Mode of the song (major or minor)
- danceability\_%:** Percentage indicating how suitable the song is for dancing
- valence\_%:** Positivity of the song's musical content
- energy\_%:** Perceived energy level of the song
- acousticness\_%:** Amount of acoustic sound in the song
- instrumentalness\_%:** Amount of instrumental content in the song
- liveness\_%:** Presence of live performance elements
- speechiness\_%:** Amount of spoken words in the song

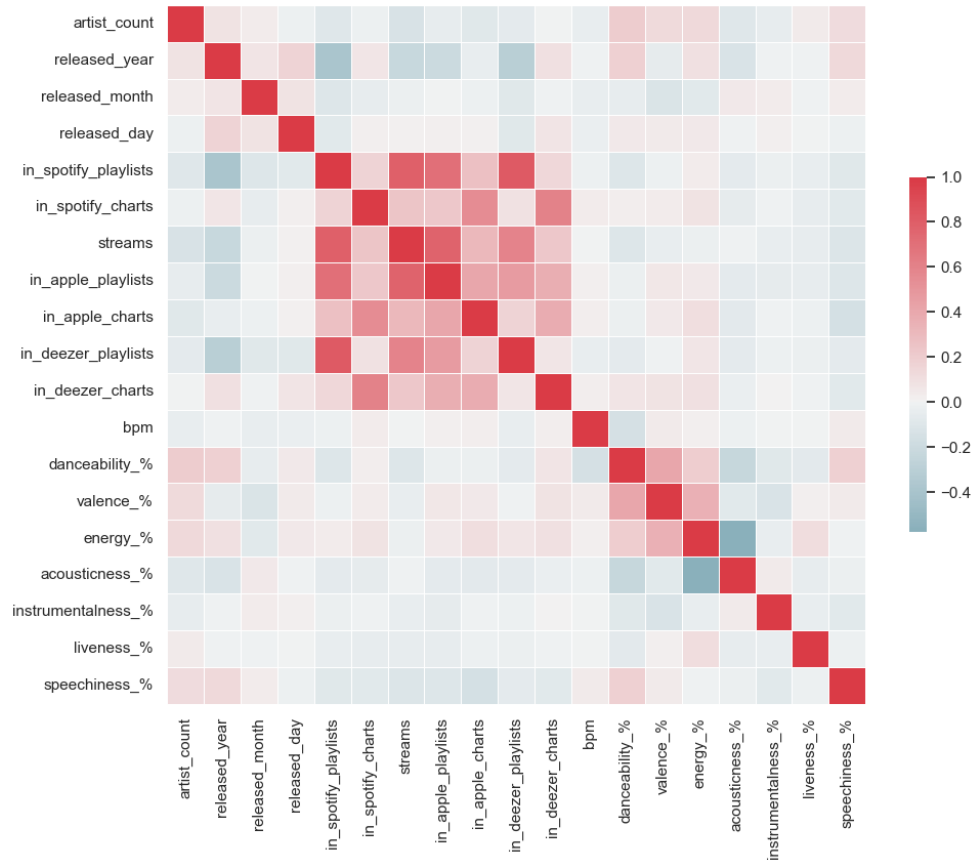
# Initial Modelling

To begin with a created a simple plot of all variables against streams to get a feel for the data.



# Modelling

This was followed by collinearity tests:



This showed that the variables with the highest collinearity with no. of streams were:

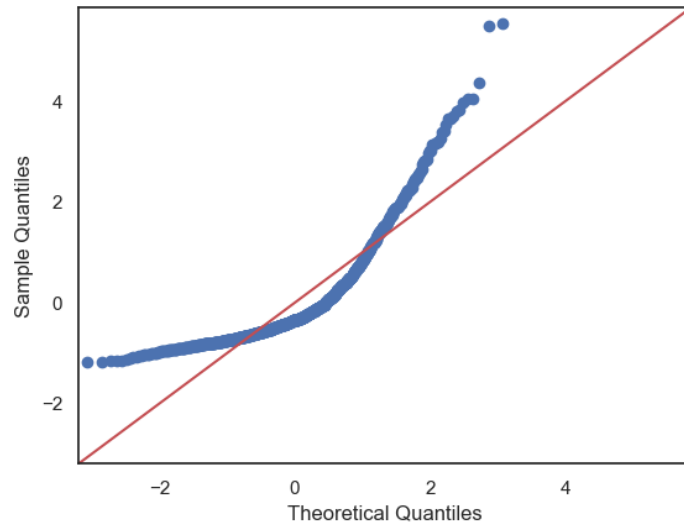
- in\_spotify\_playlists
- in\_spotify\_charts

# Modelling

To answer the brief I then reduced the dataset to just necessary variables.

Unfortunately the initial OLES results weren't favourable:

And the resulting Q-Q plot was not ideal.



## OLS Regression Results

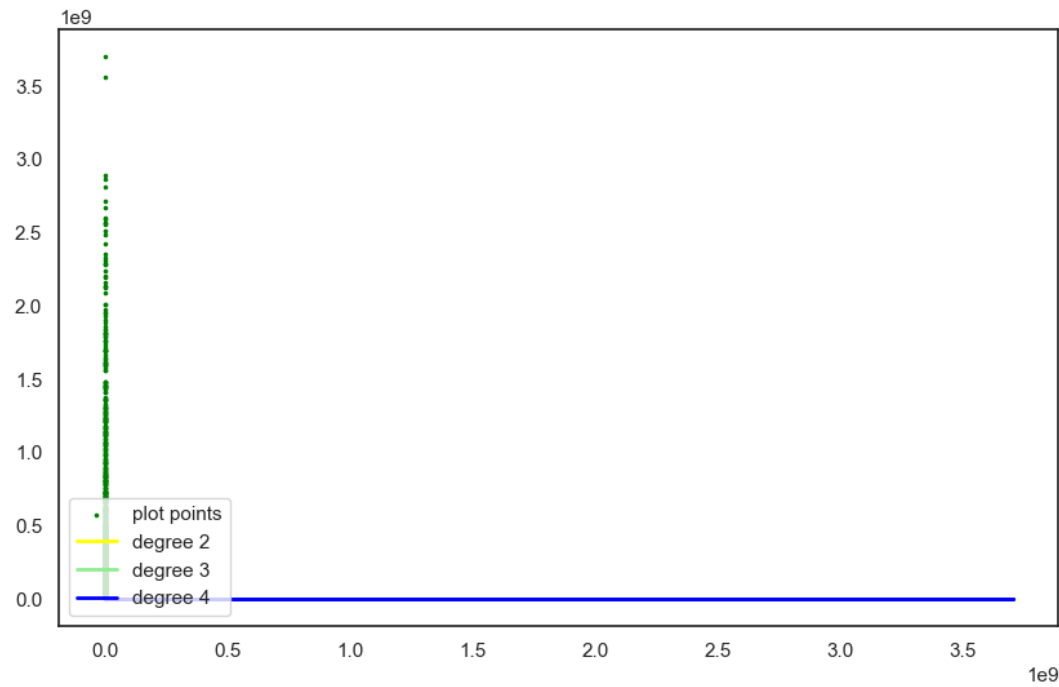
Dep. Variable:	streams	R-squared:	0.029	
Model:	OLS	Adj. R-squared:	0.021	
Method:	Least Squares	F-statistic:	3.558	
Date:	Sat, 16 Dec 2023	Prob (F-statistic):	0.000450	
Time:	12:33:15	Log-Likelihood:	-20524.	
No. Observations:	952	AIC:	4.107e+04	
Df Residuals:	943	BIC:	4.111e+04	
Df Model:	8			
Covariance Type:	nonrobust			
	coef	std err	t P> t  [0.025 0.975]	
Intercept	1.037e+09	1.69e+08	6.128 0.000	7.05e+08 1.37e+09
bpm	-3.078e+05	6.62e+05	-0.465 0.642	-1.61e+06 9.92e+05
danceability	-4.227e+06	1.46e+06	-2.886 0.004	-7.1e+06 -1.35e+06
valence	2.192e+05	9.32e+05	0.235 0.814	-1.61e+06 2.05e+06
energy	-1.119e+06	1.47e+06	-0.761 0.447	-4e+06 1.77e+06
acousticness	-1.121e+06	8.95e+05	-1.253 0.211	-2.88e+06 6.36e+05
instrumentalness	-4.291e+06	2.19e+06	-1.957 0.051	-8.59e+06 1.21e+04
liveness	-2.519e+06	1.34e+06	-1.873 0.061	-5.16e+06 1.2e+05
speechiness	-5.719e+06	1.88e+06	-3.044 0.002	-9.41e+06 -2.03e+06
Omnibus:	377.983	Durbin-Watson:	1.521	
Prob(Omnibus):	0.000	Jarque-Bera (JB):	1334.683	
Skew:	1.944	Prob(JB):	1.50e-290	
Kurtosis:	7.305	Cond. No.	1.56e+03	

# Modelling

Needing some direction I performed a stepwise selection that showed the best result would come from the two variables: speechiness and danceability.

I then proceeded create another model with only those two variables. The results did not improve. I even tried to apply polynomial regression to see if the plot could be more linear.

It was not conclusive.



# Insights

From these models I was able to conclude:

- That with this data set it is unable to be predicted whether a song would be popular from it's characteristics using a linear regression model.

From an early collinearity test I did see a high collinearity between streams and in\_spotify\_playlists. I decided to create a further model around this.

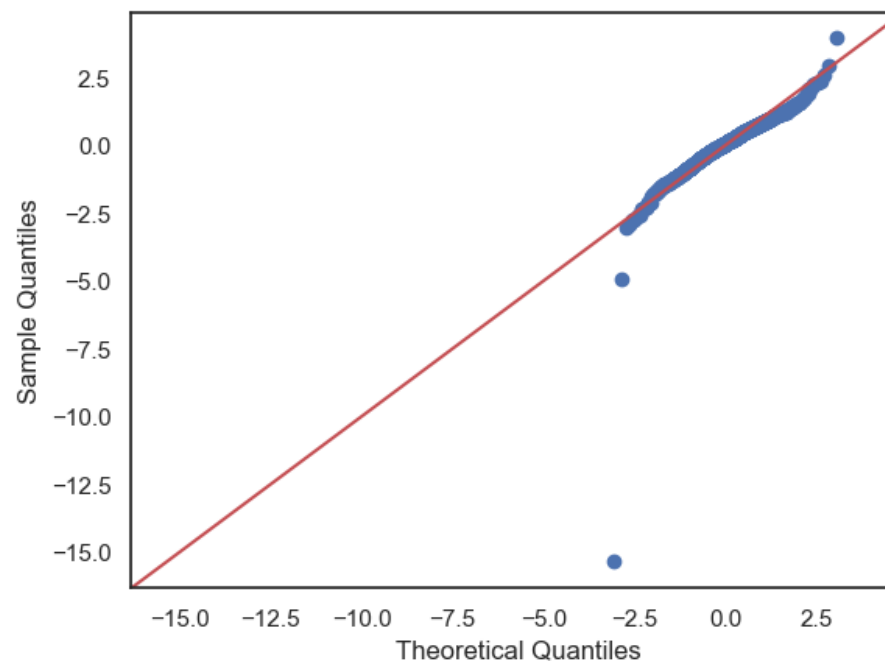


# Further Modelling.

This new model between streams and in\_spotify\_playlists had promising results:

OLS Regression Results

Dep. Variable:	streams	R-squared:	0.624	
Model:	OLS	Adj. R-squared:	0.623	
Method:	Least Squares	F-statistic:	1575.	
Date:	Sat, 16 Dec 2023	Prob (F-statistic):	6.74e-204	
Time:	13:13:24	Log-Likelihood:	-20073.	
No. Observations:	952	AIC:	4.015e+04	
Df Residuals:	950	BIC:	4.016e+04	
Df Model:	1			
Covariance Type:	nonrobust			
	coef	std err	t P> t  [0.025 0.975]	
Intercept	2.193e+08	1.35e+07	16.247 0.000	1.93e+08 2.46e+08
in_spotify_playlists	5.666e+04	1427.598	39.691 0.000	5.39e+04 5.95e+04
Omnibus:	204.656	Durbin-Watson:	1.687	
Prob(Omnibus):	0.000	Jarque-Bera (JB):	1684.085	
Skew:	0.738	Prob(JB):	0.00	
Kurtosis:	9.347	Cond. No.	1.13e+04	



# Testing

Testing was simple I split the data at 75 % to train the model and test it which resulted in:

Train Mean Squared Error:  $1.1530396750430474e+17$

Test Mean Squared Error:  $1.1945768783483576e+17$

# Conclusion

Although we may not have been able to use a linear regression model to predict the popularity of a song based on its characteristics it is possible to predict the popularity of a song based on the number of playlists it appears in, using this dataset.

# Thank you

Nyssa Mitchell

Data Scientist

[nysmitch@gmail.com](mailto:nysmitch@gmail.com)

## Any Questions?

Questions?