



Spotify:

## Relationship Between Danceability of a Song and Number of Weeks on Chart

Prae Kongchan

Ninh Nguyen

Jacquie Guerra

Van Le

Huiling Xiao


Joyce Xinyi Jiang



# Agenda: Overview


 Overview


 Hypothesis


 Ideal Experiment

 Data Preparation

 Data Visualization

 Model Explanation

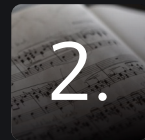
 Model Interpretation

 Conclusion

 Q&A



**1.** Introduction  
Hypothesis



**2.** Ideal  
Experimentation



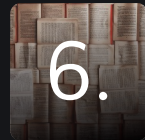
**3.** Data  
Understanding



**4.** Model  
Analysis



**5.** Conclusion  
Limitations



**6.** Q&A



# Introduction: Hypothesis



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion



Q&A

What is the similarity between “The Next Episode” by Dre Dre and “Every Breath You Take” by The Police?

Both have a dancibility score of 0.8+/1



How suitable a track is for **dancing** based on **a combination of musical elements** including tempo, rhythm stability, beat strength, and overall regularity

- Danceability > Popularity > Stays on Chart for Longer?
- By assessing each song at a more granular level by musicality, we can assess if there is something specific to each song that enables its popularity.



Null Hypothesis:  
**There is no relationship between danceability score and the length of time a song spends on chart.**



Alternative Hypothesis:  
**There is a relationship between danceability score and the length of time a song spends on chart.**



# Ideal Experiment

We will use Between-subjects (independent measures) design where songs are randomly assigned a level of danceability (none, low, or high) and follow that level of danceability throughout the experiment.



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



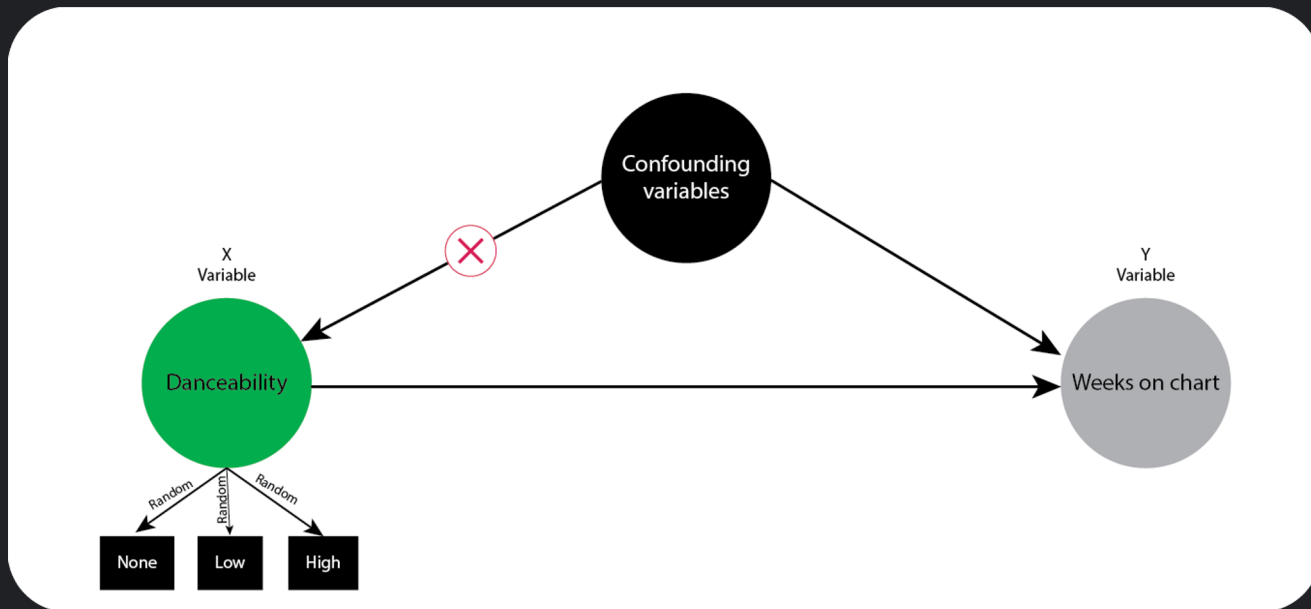
Model Interpretation



Conclusion



Q&A





# Data: Preparation

1. Combining the Datasets (using artist names and song names)
2. Create a Dummy variable for TikTok (if the song is on TikTok or not)
3. Remove Extreme Outliers
4. Transform the data (Some of the variables are right-skewed  $\rightarrow$  LOG)

Rows of Data:  
496  $\rightarrow$  447



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



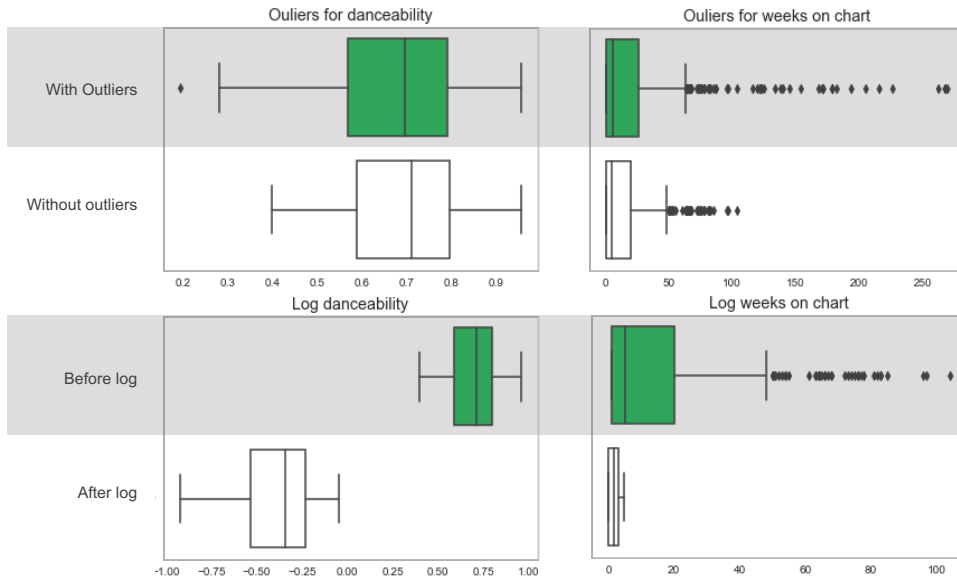
Model Interpretation



Conclusion



Q&A





# Data Description: Main Variables



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion

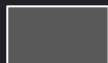


Q&A

Variables	Type	Description
LOG.weeks_on_chart	Numerical	log of number of weeks on chart
LOG.danceability	Numerical	log of how suitable a track is for dancing based on a combination of musical elements
tiktok	Categorical	1 for a song that appear on TikTok; 0 otherwise
streams	Numerical	total number of streams of the artist (in Billion)
loudness	Numerical	the quality of a sound that is the primary psychological correlate of physical strength (range: -60 to 0 db)
LOG.energy	Numerical	log of a perceptual measure of intensity and activity



Interested outcome  
dependent variables



Interested  
independent variables



control variables



# Descriptive Statistics



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion



Q&A

	Streams (Billion)	Weeks_on_chart (Week)	Danceability (0-1)	Energy (0-1)	Loudness (-60-0 Db*)
mean	16.808	14.852	0.691	0.646	-6.268
std	13.893	20.504	0.133	0.157	2.340
min	1.423	1.000	0.398	0.189	-16.169
25%	5.408	1.000	0.587	0.542	-7.454
50%	13.347	5.000	0.709	0.657	-5.883
75%	25.714	20.000	0.794	0.769	-4.604
max	50.162	104.000	0.954	0.959	-2.171



\*\* Spotify standardization metric for ideal loudness on their platform, 0 is the loudest



# Data Visualization

$r = 0.017489$

From the scatter plot, with the observational data we currently have, there is no correlation between log(week on chart) and log(danceability)



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation

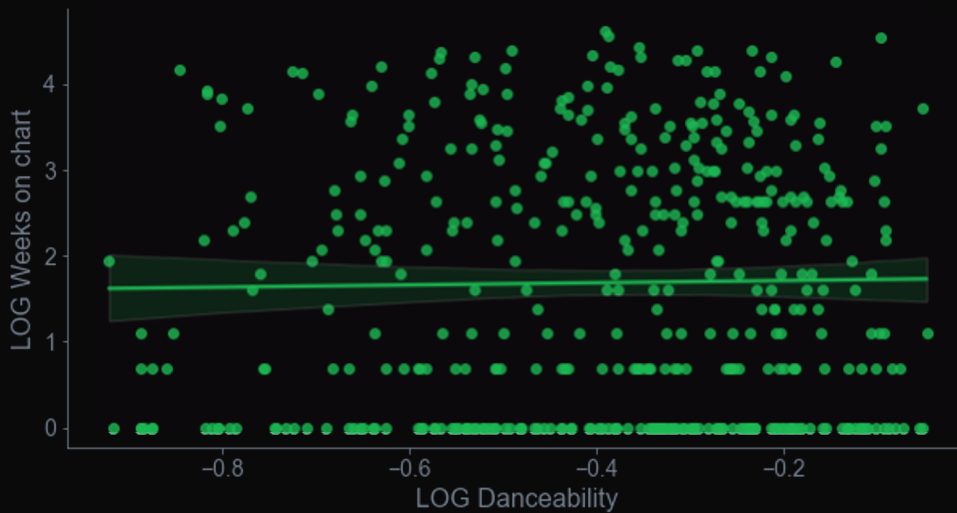


Conclusion



Q&A

A scatter plot between Danceability and Weeks on Chart







# Data Visualization: Cont.

Songs that appear on TikTok seems to have a **higher average number of weeks on chart** than songs that does not appear on TikTok



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



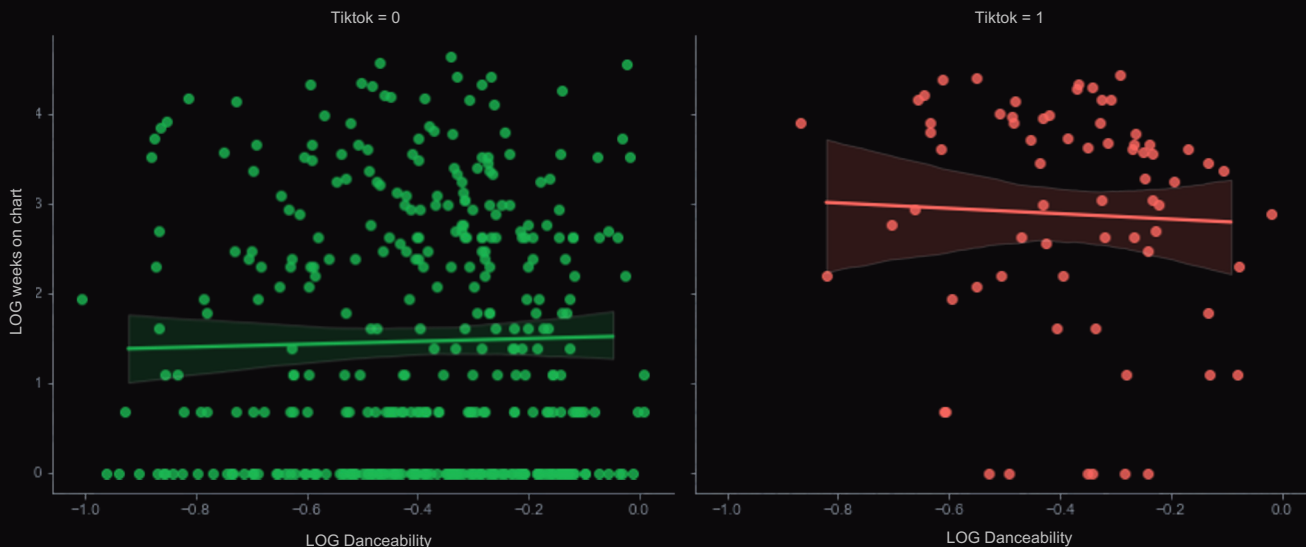
Model Interpretation



Conclusion



Q&A





# Model Explanation

$$\log(\text{weeks\_on\_chart}) = b_0 + b_1 \log(\text{danceability}) + b_2 \text{tiktok} + b_3 \text{streams} + b_4 \text{loudness} + b_5 \log(\text{energy})$$



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion



Q&A

The analysis showed that:

- $\beta_1 \approx -0.183$
- The expected weeks on chart decreases by 0.18% when danceability increases by 1%
- P-value of  $\beta_1 \approx 0.57$
- $\log(\text{danceability})$  is not significant

Residuals:

Min	1Q	Median	3Q	Max
-3.0424	-1.2265	-0.1304	1.0780	3.3988

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	1.888080	0.221429	8.527	0.000000000000000242 ***
LOG.danceability	-0.183713	0.327036	-0.562	0.57457
tiktok	1.373919	0.180577	7.609	0.000000000000169347 ***
streams	0.014845	0.004811	3.085	0.00216 **
loudness	0.128950	0.041101	3.137	0.00182 **
LOG.energy	-0.158326	0.333610	-0.475	0.63532

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.378 on 441 degrees of freedom

Multiple R-squared: 0.1607, Adjusted R-squared: 0.1512

F-statistic: 16.89 on 5 and 441 DF, p-value: 0.00000000000002748

	LOG.weeks_on_chart	LOG.danceability	tiktok	streams	loudness	LOG.energy
LOG.weeks_on_chart	1.00000000	0.01748872	0.34507451	0.08374450	0.1887917	0.12770133
LOG.danceability	0.01748872	1.00000000	0.01174009	0.02222473	0.2010677	0.16761281
tiktok	0.34507451	0.01174009	1.00000000	-0.06338938	0.1081044	0.06573248
streams	0.08374450	0.02222473	-0.06338938	1.00000000	-0.1652033	-0.03024335
loudness	0.18879175	0.20106772	0.10810436	-0.16520331	1.0000000	0.71329885
LOG.energy	0.12770133	0.16761281	0.06573248	-0.03024335	0.7132989	1.00000000



# Model Explanation: Cont.

$$\log(\text{weeks\_on\_chart}) = b_0 + b_1 \log(\text{danceability}) + b_2 \text{tiktok} + b_3 \text{streams} + b_4 \text{loudness}$$

Residuals:

Min	1Q	Median	3Q	Max
-3.0923	-1.2242	-0.1562	1.0722	3.4333

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.87931	0.22046	8.524	0.000000000000000245	***
LOG.danceability	-0.18808	0.32662	-0.576	0.5650	
tiktok	1.37481	0.18041	7.620	0.000000000000155417	***
streams	0.01456	0.00477	3.053	0.0024	**
loudness	0.11515	0.02902	3.968	0.000084573900363055	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.377 on 442 degrees of freedom

Multiple R-squared: 0.1603, Adjusted R-squared: 0.1527

F-statistic: 21.1 on 4 and 442 DF, p-value: 0.000000000000006193

The analysis showed that:

- $\beta_1 \approx -0.188$  (controlling for tiktok, streams, and loudness, the expected weeks on chart decreases by 0.18% when danceability increases by 1%)
- $\beta_2 \approx 1.375$  (295% increase in weeks on chart if TikTok = 1)
- $\beta_3 \approx 0.0145$  (1.47% increase in weeks on chart when streams increase by 1%)
- $\beta_4 \approx 0.115$  (12.2% increase in weeks on chart when loudness increases by 1%)



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion



Q&A



Overview



Hypothesis



Ideal Experiment



Data Preparation



Data Visualization



Model Explanation



Model Interpretation



Conclusion



Q&A

# Conclusion: Limitations



**We fail to reject the null hypothesis: there is no relationship between danceability and number of weeks on chart**

- > At a significance level of 0.05 and p-value of 0.5650
- > Control variables all have significant relationships: TikTok (Trends), Streams (Artist Popularity) and Loudness

**Other potential factors and limitations include:**

Seasonality

Movies/TV

Radio Play

Artist News

...etc



**Danceability is not a primary factor in music popularity, which is hard to predict in general**

Music is intertwined and collective. It is often enjoyed with a multitude of musical elements and is listened to in a social context.

Q&A

That's

#SPOTIFYWRAPPED

