

Introduce

\rightarrow What is Spotify?

Spotify is a digital music, podcast, and video service that gives you access to millions of songs and other content from creators all over the world.

\rightarrow What we did?

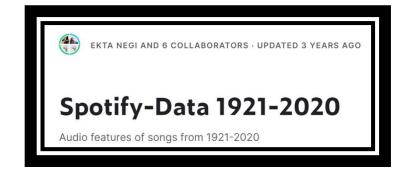
In this project, we analyzed the distribution of musical features such as danceability, energy, loudness, liveliness, valence, and duration, as well as their changes across different decades.

We also ranked the popular artists of each decade.

→ Data Source

Data in this report is extracted from <u>Kaggle-Spotify-Data 1921-2020</u>, which contains the top 100 songs in each year from 1921-2020 in Spotify (totally 169k songs) as the description.

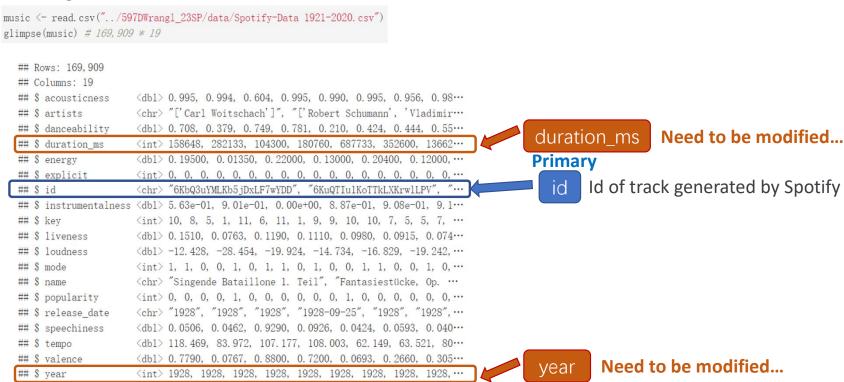




About Original Dataset

The file of dataset contains more than 160.000 songs collected from Spotify Web API. The dataset is from Spotify and contains 169k songs from the year 1921 to year 2020. Each year got top 100 songs.

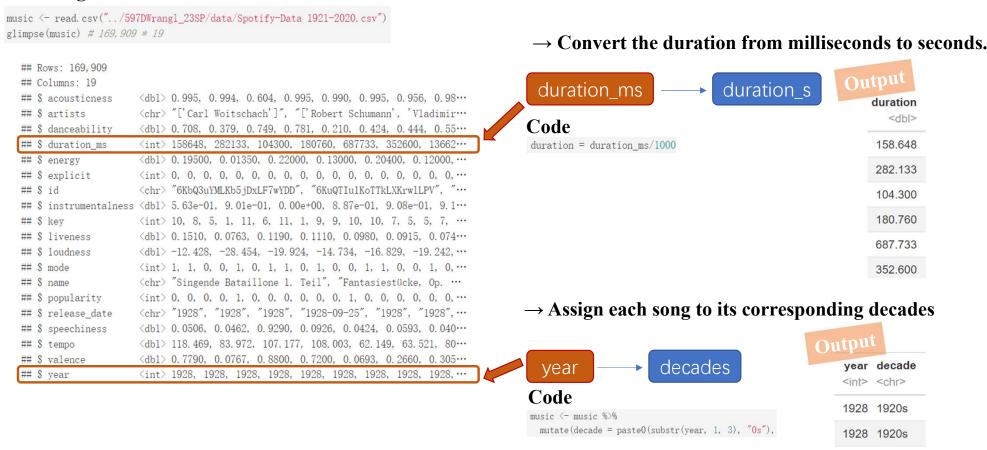
\rightarrow bring data into R



Pre-process of Data

\rightarrow bring data into R

Sentiment Analysis of Popular Songs' Lyrics



Group 16: Fangru Linghu, Yiming Tan, Siyu Chen, Arun Mishra

Clean Data and Tidy Data

\rightarrow Select variables what we need

Drop those we do not need

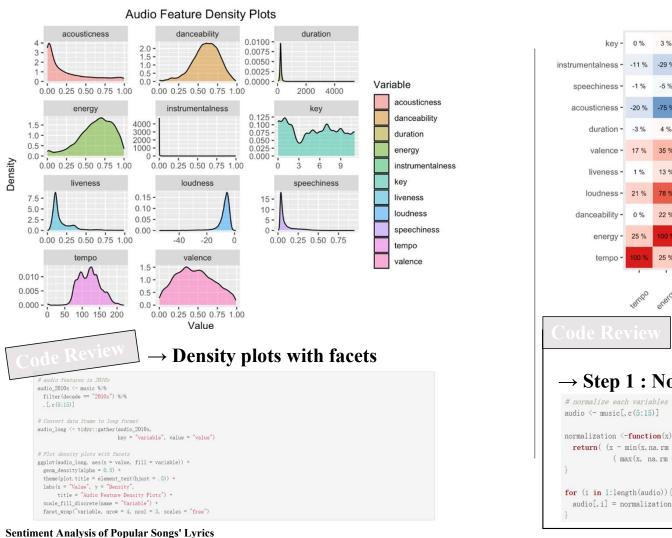
```
## Rows: 169,909
## Columns: 19
                      (db1) 0.995, 0.994, 0.604, 0.995, 0.990, 0.995, 0.956, 0.98...
## $ acousticness
## $ artists
                      <chr> "['Carl Woitschach']", "['Robert Schumann', 'Vladimir...
                     <db1> 0.708, 0.379, 0.749, 0.781, 0.210, 0.424, 0.444, 0.55...
## $ danceability
## $ duration ms
                     (int) 158648, 282133, 104300, 180760, 687733, 352600, 13662...
## $ energy
                     (db1) 0.19500, 0.01350, 0.22000, 0.13000, 0.20400, 0.12000, ...
## $ explicit
                      ## S id
                      <chr> "6KbQ3uYMLKb5jDxLF7wYDD", "6KuQTIu1KoTTkLXKrw1LPV", "...
## $ instrumentalness <dbl> 5.63e-01, 9.01e-01, 0.00e+00, 8.87e-01, 9.08e-01, 9.1...
                      (int) 10, 8, 5, 1, 11, 6, 11, 1, 9, 9, 10, 10, 7, 5, 5, 7,
## $ liveness
                      (db1) 0.1510, 0.0763, 0.1190, 0.1110, 0.0980, 0.0915, 0.074...
## $ loudness
                      <db1> -12. 428, -28. 454, -19. 924, -14. 734, -16. 829, -19. 242, ...
## 3 mode
                      <int> 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, ...
## $ name
                      <chr> "Singende Bataillone 1. Teil", "Fantasiestücke, Op. ...
                      (int) 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, ...
## $ popularity
                      (chr) "1928", "1928", "1928", "1928-09-25", "1928", "1928", ...
## $ speechiness
                      <db1> 0,0506, 0,0462, 0,9290, 0,0926, 0,0424, 0,0593, 0,040...
                      <db1> 118. 469, 83. 972, 107. 177, 108. 003, 62. 149, 63. 521, 80...
## $ tempo
## $ valence
                     (db1) 0.7790, 0.0767, 0.8800, 0.7200, 0.0693, 0.2660, 0.305...
                     (int) 1928, 1928, 1928, 1928, 1928, 1928, 1928, 1928, 1928, ...
## $ year
```

Now we have...

	name schr>				artis <chr< th=""><th></th><th></th><th></th><th></th><th></th><th>y <int></int></th></chr<>						y <int></int>
1 Singende Bataillone 1. Teil						['Carl Weitschach']					1928
2 Fantasiestücke, Op. 111; Plú tosto lento						[Robert Schumann', "Vladimir Horowitz"]					1928
3 (Chapter 1.18 -	[Sev	[Seweryn Goszczyński']					1928			
4 Bebamos Juntos - Instrumental (Romasterizado) 5 Polonaise-Fantaisie in A-Flat Major, Op. 61 6 Scherzo a capriccio: Presto						[Francisco Canaro] [Frédéric Chopin', "Vladimir Horowitz'] [Felix Mendelssohn', "Vladimir Horowitz']					1928 1928
['Carl Woitschach']						1928	1920s	118.469	0.1950		0.70
['Robert Schumann', 'Vladimir Horowitz']						1928	1920s	83.972	0.0135		0.37
[Seweryn Goszczyński]						1928	1920s	107.177	0.2200		0.74
(F	rancisco Cana	aro']				1928	1920s	108.003	0.1300		0.78
[Frédéric Chopin', 'Vladimir Horowitz']						1928	1920s	62.149	0.2040		0.21
['Felix Mendelssohn', 'Vladimir Horowitz']						1928	1920s	63.521	0.1200		0.42
•	loudness <dbl></dbl>	liveness <dbl></dbl>	valence <dbl></dbl>	duration <dbl></dbl>	acousticness <dbl></dbl>	speed	chiness <db></db>	instrume	entalness <dbl></dbl>		popularity <int< td=""></int<>
	-12.428	0.1510	0.7790	158.648	0.995		0.0506		0.563	10	
	-28.454	0.0763	0.0767	282.133	0.994		0.0462		0.901	8	9
	-19.924	0.1190	0.8800	104,300	0.604		0.9290		0.000	5	-
	-14.734	0.1110	0.7200	180.760	0.995		0.0926		0.887	- 1	- 1
	-16.829	0.0980	0.0693	687,733	0.990		0.0424		0.908	11	- 2
	-19.242	0.0915	0.2660	352.600	0.995		0.0593		0.911	6	- 1

Summary of the data...

Comparison among variables and Relationship between each two variables



3% 2% 2% 0% 3% 0% -2% 23 % -11 % -2 % correlations 1.0 -13 % -1 % -8 % 0 % 0.5 -20 % -19 % 56 % 31 % 0 % -19 % 3 % 0.0 13 % -11 % 5 % 0 % -2 % 15 % -5 % 0 % -0.5 5 % -57 % -11 % -42 % 29 % 2 % -10 29 % -13 % -27 % 23 % -29 % 3 % \rightarrow Step 2 : tile plot cor() %>% melt() %>% \rightarrow Step 1 : Normalizing ggplot(aes(X1, X2, fill=value)) + geom_tile(color = 'white') + scale_fill_gradient2(low = "#2C7BB6", mid = "white", high = "#D7191C", midpoint = 0, name = "correlations", limits = c(-1, 1), na. value = "gray90", guide = "colorbar", normalization (-function(x) oob = scales::squish) + return((x - min(x, na. rm = T)))/geom text(aes(label = paste(round(value, 2) * 100, '%')), $(\max(x, na.rm = T) - \min(x, na.rm = T)))$ size = 2.5, color = 'black') + labs (x = '', y = '')title = 'Relationship between music features') + theme (axis. text. x = element_text (angle = 45, vjust = 0.5), for (i in 1:length(audio)) { plot.title = element_text(hjust = 0.5), audio[, i] = normalization(audio[, i]) panel.grid.major = element_blank(), panel.grid.minor = element_blank())

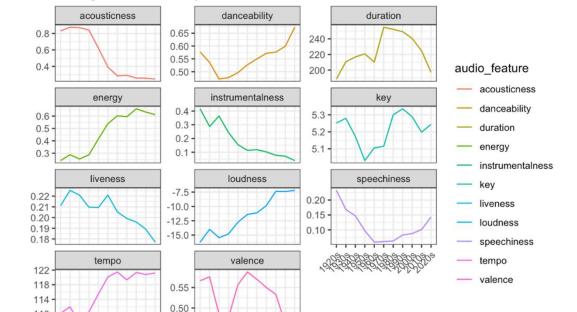
Group 16: Fangru Linghu, Yiming Tan, Siyu Chen, Arun Mishra

Relationship between music features

Changes from 1921 to 2020

```
Code Review → Line plots
```

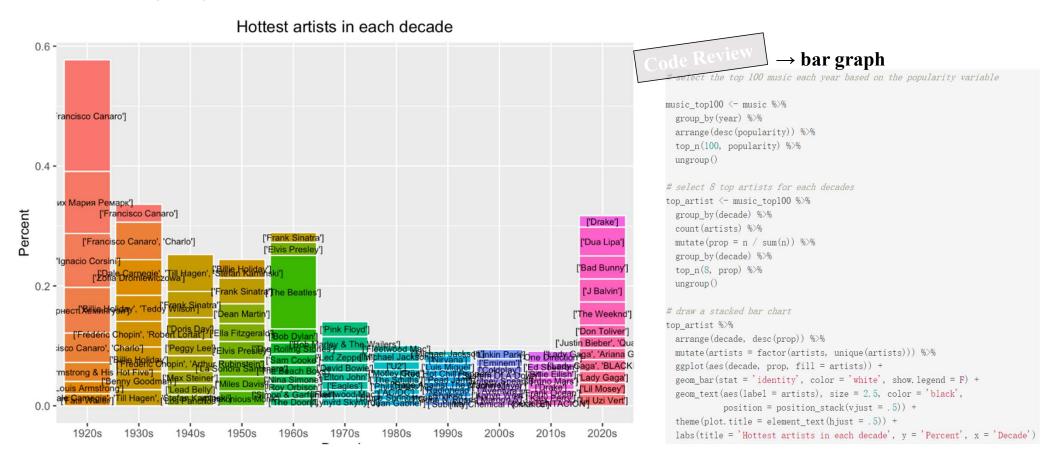
```
audio_decade <- music[, c(4, 5:15)]
audio decade %>%
 group by (decade) %>%
 summarize (across (starts with ("tempo"), mean),
           across(starts_with("energy"), mean),
           across(starts_with("danceability"), mean),
           across(starts with("loudness"), mean),
           across(starts with("liveness"), mean),
           across(starts_with("valence"), mean),
           across(starts_with("duration"), mean),
           across(starts with("acoustioness"), mean),
           across(starts_with("speechiness"), mean),
           across(starts with("instrumentalness"), mean),
           across(starts with("key"), mean)) %>%
 pivot_longer(cols = -decade, names_to = "audio_feature",
              values_to = "value") %>%
 ggplot(aes(x = decade, y = value, group = audio feature,
             color = audio_feature)) +
 geom line() +
 facet_wrap(~ audio_feature, nrow = 4, ncol = 3,
             scales = "free v") +
 labs(title = "Change in Features by Decade") +
 theme bw() +
 theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



decade

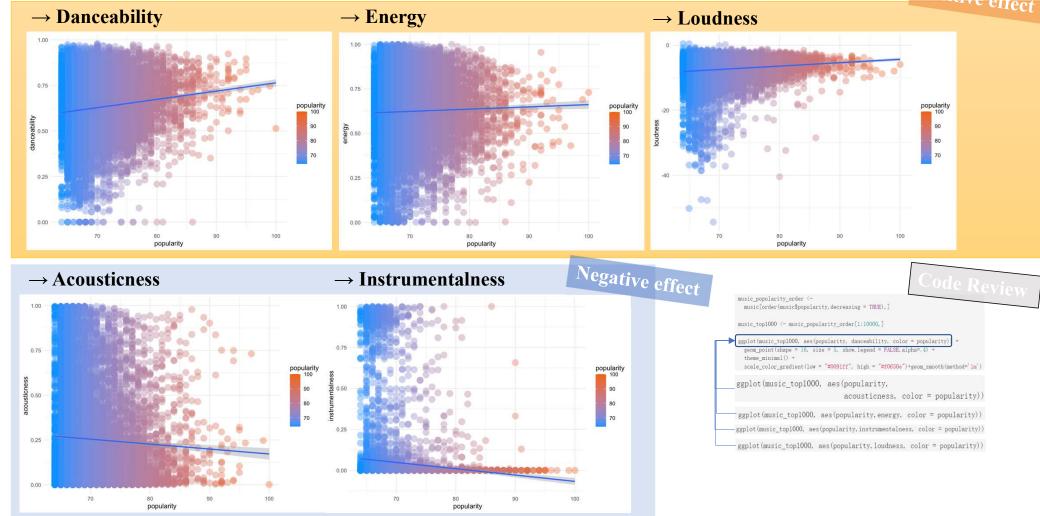
Change in Features by Decade

Rank of popular artists of each decade



Attributes vs. Popularity

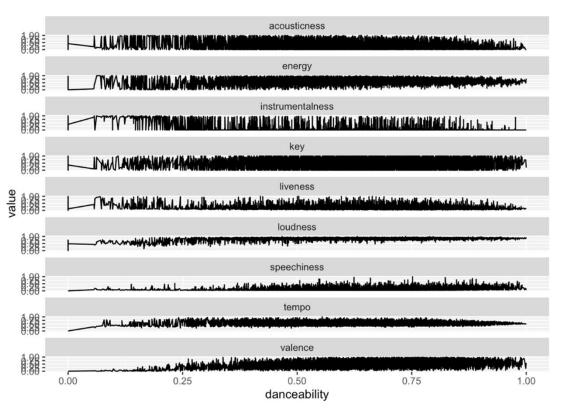




Sentiment Analysis of Popular Songs' Lyrics

Group 16: Fangru Linghu, Yiming Tan, Siyu Chen, Arun Mishra

What Features are important for a song to be danceable?



```
# function to scale the values between 0 and 1
regularization <- function(x) {
   (x - min(x)) / (max(x) - min(x))
}

danceable_music <- music_top1000 %>%
   arrange(desc(popularity)) %>%
   select(-c(name, artists, year, decade, popularity, duration)) %>%
   mutate(across(everything(), regularization)) %>%
   pivot_longer(cols = -c(danceability), names_to = "variable", values_to = "value",

ggplot(danceable_music, aes(danceability, value)) +
   geom_line() +
   facet_wrap(`variable, scales = "free_y", ncol = 1)
```

Most common words in title with a WordCloud

in 21st Century:



Summary

Common Characteristics of Popular Songs

- lower acousticness
- higher energy
- > less instrumentation
- higher loudness
- higher danceability

Significance of Project

- These findings can help understand the changing trends in music over the years and the key factors contributing to a song's popularity.
- However, it is essential to note that this analysis is based on a limited dataset and does not consider the complete landscape of the music industry.

Thank you for Listening!