# McDonald's Sentiment Analysis

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# **Agenda**

**Industry Background** 

**Data Preparation** 

Perceptual map

Strategic Insights



# **Industry Background**



### **Industry and Business Background**

### **Industry Overview**

#### **Key Statistics:**







202K businesses

**1.8%** CAGR

**\$18.7B** Profit

#### **Major Players:**







### **Business Background**

- McDonald's plans to open 1,900 new locations in 2023
- McDonald's anticipate over 4% unit growth from net restaurant additions in 2023

### **Project Objectives**

- Identify, analyze, and categorize online sentiments about Mcdonald's
- Perform sentiment analysis and produce a perceptual map that captures brands delights and disappointment
- Provide insights and recommendations to McDonald's based on sentiment analysis



# **Data Preparation**



### **Data Preparation**

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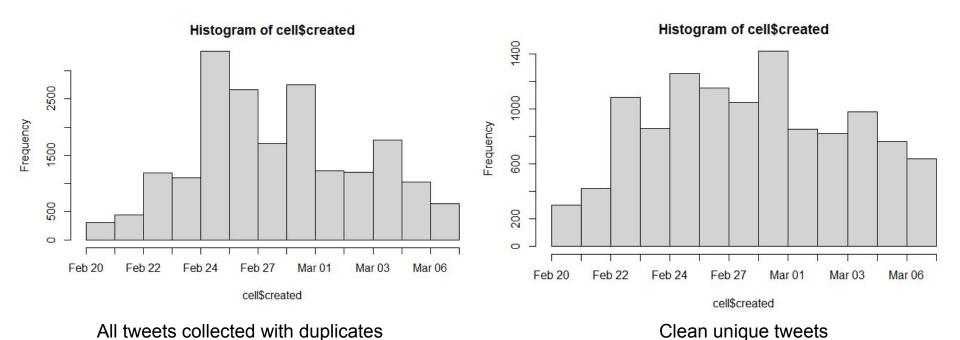
- Pull online sentiment twitter data for each of the brands in 2023
- Twitter API can only return portion of data (around 500 per brand per day), we merged data from several days to increase to volume of the data
- Data cleaning: remove stop words, punctuations, numbers, and converted all alphabets to lowercase
- We only keep McDonald's, KFC, Starbucks, and Subway for analysis due to data limitations

### **Data Breakdown by Tweets**

Brands	Number of Tweets
McDonald's	3.6k
KFC	2.3k
Starbucks	4.3k
Subway	1.5k
Chipotle*	0.5k
Chick-fil-a*	0.4k



# **Time-Range of Tweets**





### **Methodology and Mapping Processes**

**Step 1: Data Cleaning all the tweets** 

Step 2: Using Latent Dirichlet Allocation (LDA) to find the topics and label attributes

Step 3: Using keywords to find tweets related to topics

Step 4: Perform sentiment analysis on the tweets per topic and brand

Step 5: Create a Perceptual map using Multi-dimensional scaling of brand sentiments per attribute



# **Perceptual Map**



### **Latent Dirichlet Allocation**

topic_num <int></int>	topic_size <dbl></dbl>	mean_token_length	dist_from_corpus	tf_df_dist <dbl></dbl>	doc_prominence	topic_coherence	topic_exclusivity <dbl></dbl>
1	2388.792	4.7	0.6826975	6.651704	417	-242.283290	9,936054
2	1702.856	6.7	0.6862012	8.222201	1048	-204.282406	9.957862
3	2524.081	4.1	0.6841852	2.850474	376	-245.766896	9.852280
4	1254.778	6.6	0.6791808	16.274681	1085	-8.076977	9.997783
5	2379.642	4.4	0.6810718	4.989780	382	-254.561788	9.864533
6	2042.284	9.1	0.6839816	4.885628	592	-90.443903	9.941889
7	2419.346	6.7	0.6870836	1.128117	484	-288.612218	9.926344
8	2472.188	4.1	0.6922855	1.273543	382	-234.576570	9.993408
9	2470.568	4.8	0.6873543	4.276997	285	-248.901451	9.936762
10	1316.464	4.1	0.6661678	11.425094	770	-9.869056	9.996113

Topic size: measures size of topic

Doc prominence: measures how prominent a topic is

Topic coherence: measures semantic similarity

Topic 4: Brand reputation

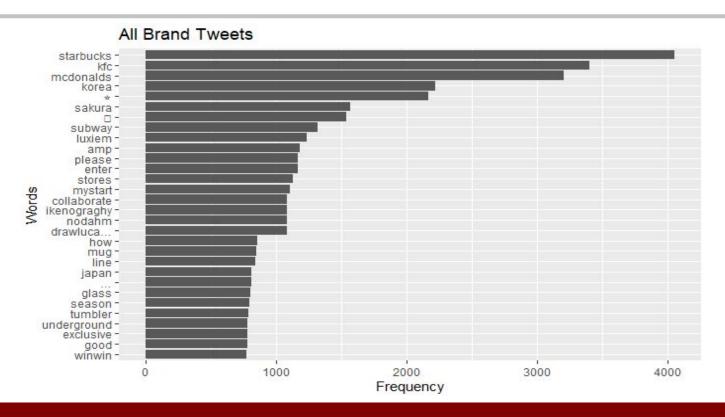
Topic 6: Advertising and Promotions

Topic 7: Food and Beverages

Topic 10: Social Media Engagement

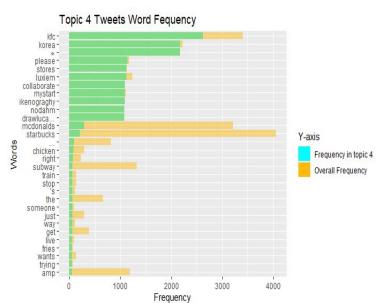


### **Overall Frequency of Words in Tweets**



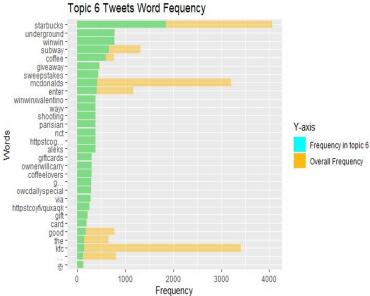
### Word Frequencies - Topic 4 & Topic 6

### **Brand Reputation**



korea, luxiem, please, stores, collaborate, customerservice, fraud, delivered, fake, eats

### **Advertising and Promotions**

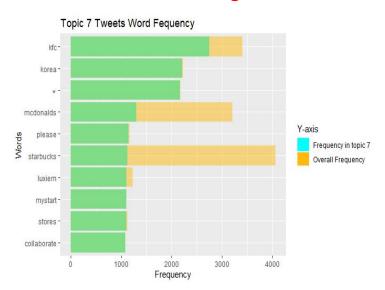


giveaway, gift, card, good, dailyspecial, winwin, value, platform, extra, foodreview, cup, elevated



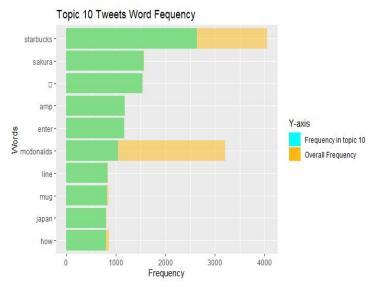
### Word Frequencies - Topic 7 & Topic 10

#### **Food and Beverages**



food, fastfood, drinks, happymeal, restaurant, frappuccino, vanilla, party

### **Social Media Engagement**

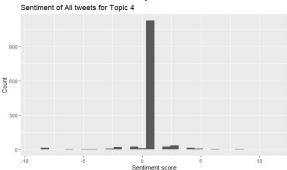


sakura, japan, exclusive, shining, giveaway, biggest, voice, apologize, england, enter

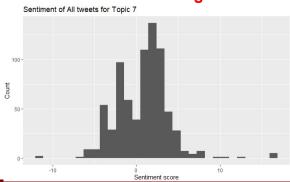


### Overall sentiments for selected topics

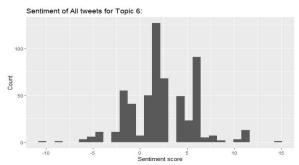
#### **Brand Reputation**



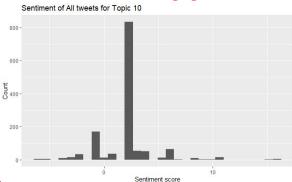
#### Food and Beverages



#### **Advertising and Promotions**



### **Social Media Engagement**





### Brand sentiment scores for each brand

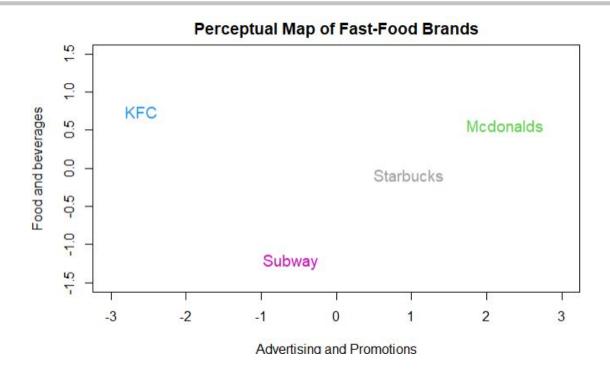
	Brand Reputation	Advertising and Promotions	Food and Beverages	Social Media Engagement
Mcdonalds	0.2195122	3.4193550	0.8156028	1.015428
KFC	0.9682647	-1.3076920	1.3809520	1.622642
Subway	2.3333330	0.8888889	0.3132530	1.523810
Starbucks	1.1666670	2.3800740	1.1992750	2.061545

Applying Multi-dimensional Scaling:

Brand	X1	X2
Mcdonalds	2.2450407	0.56099680
KFC	-2.5950542	0.74430033
Subway	-0.6146597	-1.22281011
Starbucks	0.9646732	-0.08248702



### **Perceptual Map**



# **Strategic Insights**



### Strategic Insights

### **Takeaways**

#### **Competitors:**

- Subway has not been doing well lately. Currently being sold and there are already 3 bidders. The company fell in value from \$16B in 2016 to \$10B in 2022
- KFC has the overall best food and beverages, but suffers in advertising and promotions. They try to make up for this with their social media engagement.
- Starbucks performs well across all attributes. Their social media engagement is the best and they have regular promotions and giveaways

#### Mcdonald's:

- Mcdonald's has the best advertising and Promotions which makes up for its average food and beverage attributes.
- Mcdonald's could put more effort in improving food quality to increase customer perception
- Mcdonald's is the worst brand for Social Media Engagement. We need to take advantage of our Advertising and Promotions to further engagement with customers.



### **Limitations and Future Improvements**

#### Limitations

- We could only pull a maximum of about 2000 tweets per day per brand. The twitter API only returns a few of the tweets daily and we merged the tweets for each brand daily to collect sufficient unique tweets per brand
- Data is only pulled from twitter. Other channels like news articles, other social media platforms and reviews website could be used for a more comprehensive analysis

### **Future Improvements**

- Gather more data on different social media platforms, include more competitors in the perceptual map, and generate topics like customer service
- Perform sentiment analysis in different domains, for example, in different countries and regions