

Introduction

Hello, I am Nelly and I am in the final stages of completing my Google Data Analytics certificate. This project thoroughly analyzes Social Buzz content categories and highlights the top 5 categories with the largest aggregate popularity. I will apply "Ask, Prepare, Process, Analyze, Share, and Act" methodology effectively throughout the data analytics process. [Contact Me](#)

Ask phase

Social-Buzz, a rapidly expanding media content creation company that prioritizes user anonymity and monitors user reactions on every piece of content.

The company is currently facing challenges related to managing its expanding scale. Over 100,000 pieces of content, ranging from text, images, videos, and GIFs are posted every day. Accenture has begun a 3-month project to address key areas, including an audit of Social Buzz's big data practice, recommendations for a successful IPO, and an analysis to identify the company's top 5 most popular content categories. Given the substantial amount of data available, Social Buzz needs an effective strategy to leverage this information and maximize its potential for growth.

This project thoroughly analyzes their content categories and highlights the top 5 categories with the largest aggregate popularity.

Prepare phase

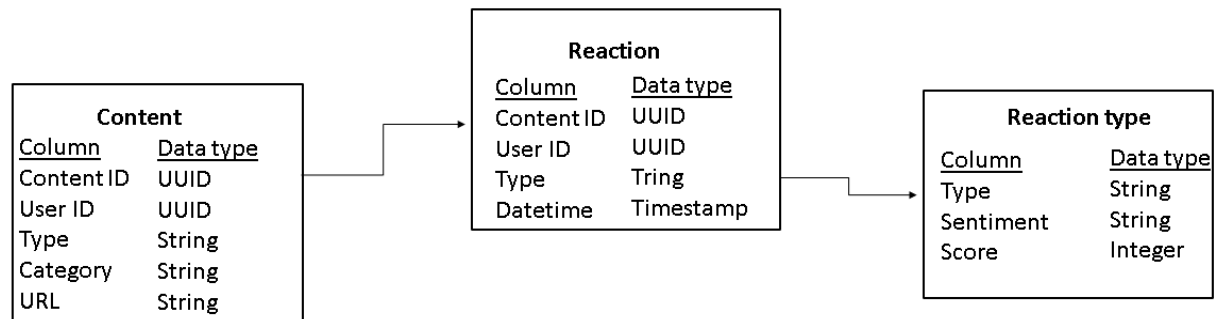
Data source: these datasets were provided by Accenture and stored on [forage](#). Forage is a free virtual work experience program from top companies.

The dataset consists of a sample of over 25,000 reactions, 16 distinct qualitative content, and a quantitative score that is directly related to the sentiment and type of the reaction types.

Accenture supplied three CSV files (Reactions, Reaction type, and Content). Given the small sample size, I will utilize Excel to complete this project. To be accessible, I have merged the three tables into a single spreadsheet. Merging the tables was not essential for this step but it enables me to present all three data sets in a unified format.

	A	B	C	D	E	F	G
1		Content ID	User ID	Type	Datetime		
2	0	97522e57-d9ab-4bd6-97bf-c24d952602d2			4/22/2021 15:17		
3	1	97522e57-d9ab-4bd6-97bf-c24d952602d2	5d454588-283d-459d-915d-c48a2cb4c27f	disgust	11/7/2020 9:43		
4	2	97522e57-d9ab-4bd6-97bf-c24d952602d2	92b87fa5-f271-43e0-af66-84fac21052e6	dislike	6/17/2021 12:22		
5	3	97522e57-d9ab-4bd6-97bf-c24d952602d2	163daa38-8b77-48c9-9af6-37a6c1447ac2	scared	4/18/2021 5:13		
5	4	97522e57-d9ab-4bd6-97bf-c24d952602d2	34e8add9-0206-47fd-a501-037b994650a2	disgust	1/6/2021 19:13		
7	5	97522e57-d9ab-4bd6-97bf-c24d952602d2	9b6d35f9-5e15-4cd0-a8d7-b1f3340e02c4	interested	8/23/2020 12:25		
8	6	97522e57-d9ab-4bd6-97bf-c24d952602d2	7918d465-0953-4f20-9e28-539e74c82e2f	peeking	12/7/2020 6:27		
9	7	97522e57-d9ab-4bd6-97bf-c24d952602d2	fa3e42f0-71d8-455f-b024-e52d5c27a145	cherish	4/11/2021 17:35		
0	8	97522e57-d9ab-4bd6-97bf-c24d952602d2	b0c22f82-b882-4394-bf27-6dfadf26e5c2	hate	1/27/2021 8:32		
1	9	97522e57-d9ab-4bd6-97bf-c24d952602d2	1932a904-86ba-4438-bb52-b7e6516a4019	peeking	4/1/2021 22:54		
2	10	97522e57-d9ab-4bd6-97bf-c24d952602d2		love	8/4/2020 5:05		
3	11	97522e57-d9ab-4bd6-97bf-c24d952602d2	f50ac030-3af8-4e07-aacf-dccff353b8f6	indifferent	11/7/2020 8:36		
4	12	97522e57-d9ab-4bd6-97bf-c24d952602d2	b09e3125-2162-443b-bea0-25895b7605bc	scared	11/2/2020 6:28		
5	13	97522e57-d9ab-4bd6-97bf-c24d952602d2	b143cd2b-20b1-4e9c-9bb7-cfb359b33d01	interested	11/1/2020 1:16		
6	14	97522e57-d9ab-4bd6-97bf-c24d952602d2	48e834aa-fd4a-4206-93e8-ac8fdac613bc	hate	10/7/2020 18:39		
7	15	97522e57-d9ab-4bd6-97bf-c24d952602d2	7246fed1-8725-4d6b-bccd-5451346f122d	scared	9/3/2020 18:51		
8	16	97522e57-d9ab-4bd6-97bf-c24d952602d2	9664d628-9163-4aae-94e0-311f3bdb2f11	super love	2/24/2021 5:09		

Data model



Content

ID: Unique ID of the content that was uploaded (automatically generated)

User ID: Unique ID of a user that exists in the User table

Type: A string detailing the type of content that was uploaded

Category: A string detailing the category that this content is relevant to

URL: Link to the location where this content is stored

Reaction

Content ID: Unique ID of a piece of content that was uploaded

User ID: Unique ID of a user that exists in the User table who reacted to this piece of content

Type: A string detailing the type of reaction this user gave

Datetime: The date and time of this reaction

ReactionTypes

Type: A string detailing the type of reaction this user gave

Sentiment: A string detailing whether this type of reaction is considered as positive, negative or neutral

Score: This is a number calculated by Social Buzz that quantifies how “popular” each reaction is. A reaction type with a higher score should be considered as a more popular reaction.

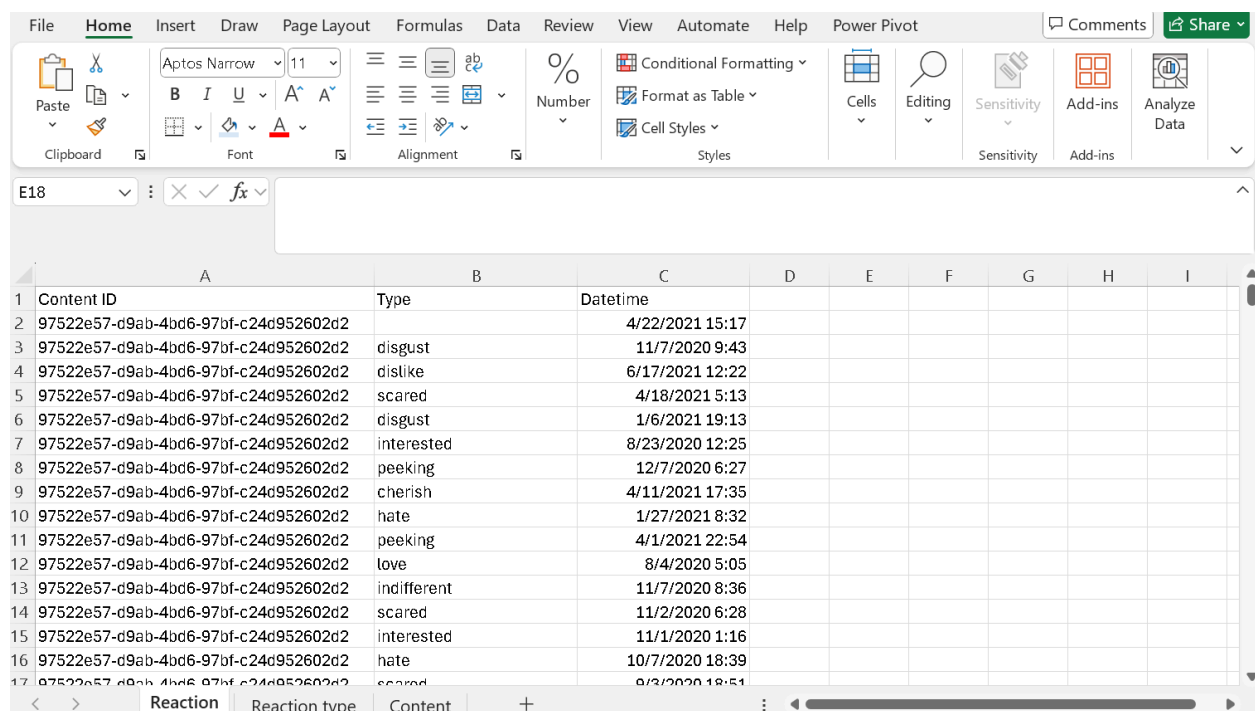
Process phase

Removing columns which are not relevant to this task

Content: removed User ID and URL column

Reaction sheet: removed User ID column

Lastly, I removed the first column (numbered column) of all three data because Excel already counts rows.



	A	B	C	D	E	F	G	H	I
1	Content ID	Type	Datetime						
2	97522e57-d9ab-4bd6-97bf-c24d952602d2		4/22/2021 15:17						
3	97522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	11/7/2020 9:43						
4	97522e57-d9ab-4bd6-97bf-c24d952602d2	dislike	6/17/2021 12:22						
5	97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	4/18/2021 5:13						
6	97522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	1/6/2021 19:13						
7	97522e57-d9ab-4bd6-97bf-c24d952602d2	interested	8/23/2020 12:25						
8	97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	12/7/2020 6:27						
9	97522e57-d9ab-4bd6-97bf-c24d952602d2	cherish	4/11/2021 17:35						
10	97522e57-d9ab-4bd6-97bf-c24d952602d2	hate	1/27/2021 8:32						
11	97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	4/1/2021 22:54						
12	97522e57-d9ab-4bd6-97bf-c24d952602d2	love	8/4/2020 5:05						
13	97522e57-d9ab-4bd6-97bf-c24d952602d2	indifferent	11/7/2020 8:36						
14	97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	11/2/2020 6:28						
15	97522e57-d9ab-4bd6-97bf-c24d952602d2	interested	11/1/2020 1:16						
16	97522e57-d9ab-4bd6-97bf-c24d952602d2	hate	10/7/2020 18:39						
17	97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	9/3/2020 18:51						

Removing rows that have values which are missing, duplicate, and data extraction

I applied filters on all three sheets separately (Reaction, Reaction type, and Content) to inspect for duplicate, blanks, and errors (typo).

Reaction sheet

- Extracted month from datetime column by using the convert text to column wizard and renamed the columns (Month, Time).
- Joined the time column by using the formula " =TEXT(cell, "hh:mm") & " " & cell" and changed data type to display month in "text".
- Changed the month column to "short date" and time to "time"(Ref R).

Content ID	Reaction Type	Month	Time
37522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	November	09:43 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	dislike	June	12:22 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	scared	April	05:13 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	January	07:13 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	interested	August	12:25 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	December	06:27 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	cherish	April	05:35 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	hate	January	08:32 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	April	10:54 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	love	August	05:05 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	indifferent	November	08:36 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	scared	November	06:28 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	interested	November	01:16 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	hate	October	06:39 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	scared	September	06:51 PM
37522e57-d9ab-4bd6-97bf-c24d952602d2	super love	February	05:09 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	September	06:24 AM
37522e57-d9ab-4bd6-97bf-c24d952602d2	indifferent	February	11:37 AM

Ref R: Month extraction

Reaction type sheet

- Fixed the score duplicate adding "1" to the want type: new score 71 (Ref Rt)

Type	Sentiment	Score
want	positive	71
cherish	positive	70

Ref Rt: Duplicate

Content sheet

- In the category column, I used the “LOWER” function to change values to lowercase. (In an empty column type “=LOWER(cell)”, use the fill handle to copy the formula. Lastly, I only pasted the values (*Ref C1*).
- I used the find and replace quotation “ ” with blank (*Ref C2*).
- I replaced “dogs” with animals (*Ref C3*).

File

Home

Insert

Draw

Page Layout

Formulas

Data

Review

View

Automate

Help

Power Pivot

Comments

Share

Paste

Clipboard

B

I

U

A[^]

A^v

Font

Alignment

Number

Conditional Formatting

Format as Table

Cell Styles

Styles

Cells

Editing

Sensitivity

Sensitivity

Add-ins

Add-ins

Analyze Data

D3

Find and Replace

Find Replace

Find what: " No Format Set Format...

Replace with: No Format Set Format...

Within: Sheet ☐ Match case

Search: By Columns ☐ Match entire cell contents

Look in: Formulas Options <<

Replace All Replace Find All **Find Next** Close

Book	Sheet	Name	Cell	Value	Formula
Content.csv	Content		\$C\$52	"culture"	
Content.csv	Content		\$C\$102	"studying"	
Content.csv	Content		\$C\$152	"animals"	
Content.csv	Content		\$C\$202	"soccer"	

17 cell(s) found

Ref C2: Removing quotation mark

Find and Replace

Find Replace

Find what: dogs No Format Set Format...

Replace with: animals No Format Set Format...

Within: Sheet ☐ Match case

Search: By Columns ☐ Match entire cell contents

Look in: Formulas Options <<

Replace All Replace Find All **Find Next** Close

Book	Sheet	Name	Cell	Value	Formula
Content.csv	Content		\$C\$28	dogs	
Content.csv	Content		\$C\$33	dogs	
Content.csv	Content		\$C\$46	dogs	
Content.csv	Content		\$C\$61	dogs	

58 cell(s) found

(Ref C2): Replaced "dogs" with animals

Aggregation

I used the Reaction table as the base table, then first joined the relevant columns from the Content data set, and then the Reaction Types data set. I named the tables to enable the use of the “VLookUp” formula, and renamed the joined column (Ref Ag).

Content ID	Reaction Type	Month	Time	Content Type	Content Category	Sentiment	Score
97522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	November	09:43 AM	photo	studying	negative	0
97522e57-d9ab-4bd6-97bf-c24d952602d2	dislike	June	12:22 PM	photo	studying	negative	10
97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	April	05:13 AM	photo	studying	negative	15
97522e57-d9ab-4bd6-97bf-c24d952602d2	disgust	January	07:13 PM	photo	studying	negative	0
97522e57-d9ab-4bd6-97bf-c24d952602d2	interested	August	12:25 PM	photo	studying	positive	30
97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	December	06:27 AM	photo	studying	neutral	35
97522e57-d9ab-4bd6-97bf-c24d952602d2	cherish	April	05:35 PM	photo	studying	positive	70
97522e57-d9ab-4bd6-97bf-c24d952602d2	hate	January	08:32 AM	photo	studying	negative	5
97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	April	10:54 PM	photo	studying	neutral	35
97522e57-d9ab-4bd6-97bf-c24d952602d2	love	August	05:05 AM	photo	studying	positive	65
97522e57-d9ab-4bd6-97bf-c24d952602d2	indifferent	November	08:36 AM	photo	studying	neutral	20
97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	November	06:28 AM	photo	studying	negative	15
97522e57-d9ab-4bd6-97bf-c24d952602d2	interested	November	01:16 AM	photo	studying	positive	30
97522e57-d9ab-4bd6-97bf-c24d952602d2	hate	October	06:39 PM	photo	studying	negative	5
97522e57-d9ab-4bd6-97bf-c24d952602d2	scared	September	06:51 PM	photo	studying	negative	15
97522e57-d9ab-4bd6-97bf-c24d952602d2	super love	February	05:09 AM	photo	studying	positive	75
97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	September	06:24 AM	photo	studying	neutral	35
97522e57-d9ab-4bd6-97bf-c24d952602d2	indifferent	February	11:37 AM	photo	studying	neutral	20
97522e57-d9ab-4bd6-97bf-c24d952602d2	interested	May	07:44 PM	photo	studying	positive	30
97522e57-d9ab-4bd6-97bf-c24d952602d2	intrigued	January	04:03 PM	photo	studying	positive	45
97522e57-d9ab-4bd6-97bf-c24d952602d2	peeking	November	05:26 PM	photo	studying	neutral	35
97522e57-d9ab-4bd6-97bf-c24d952602d2	worried	April	08:47 PM	photo	studying	negative	12
97522e57-d9ab-4bd6-97bf-c24d952602d2	like	June	04:46 PM	photo	studying	positive	50
97522e57-d9ab-4bd6-97bf-c24d952602d2	heart	April	02:29 PM	photo	studying	positive	60
97522e57-d9ab-4bd6-97bf-c24d952602d2	worried	March	07:21 PM	photo	studying	negative	12

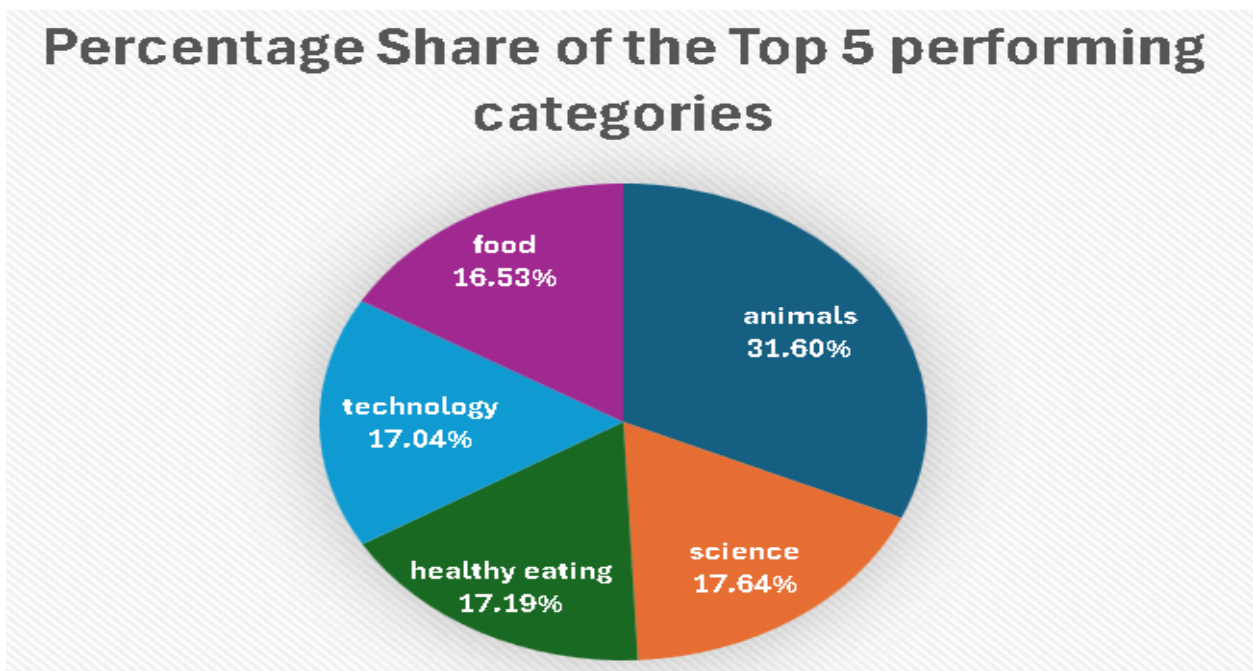
(Ref Ag): Aggregation

Analyze phase

Figure out the Top 5 performing categories by adding up the total scores for each category (using the “Sum If” formula). Next, I sorted the data by “sum of score” DESC.

Top 5 performing categories	
Content Category	Sum of score
animals	127681
science	71294
healthy eating	69450
technology	68851
food	66786

Next, I used a pie chart to represent the **top 5 categories**.

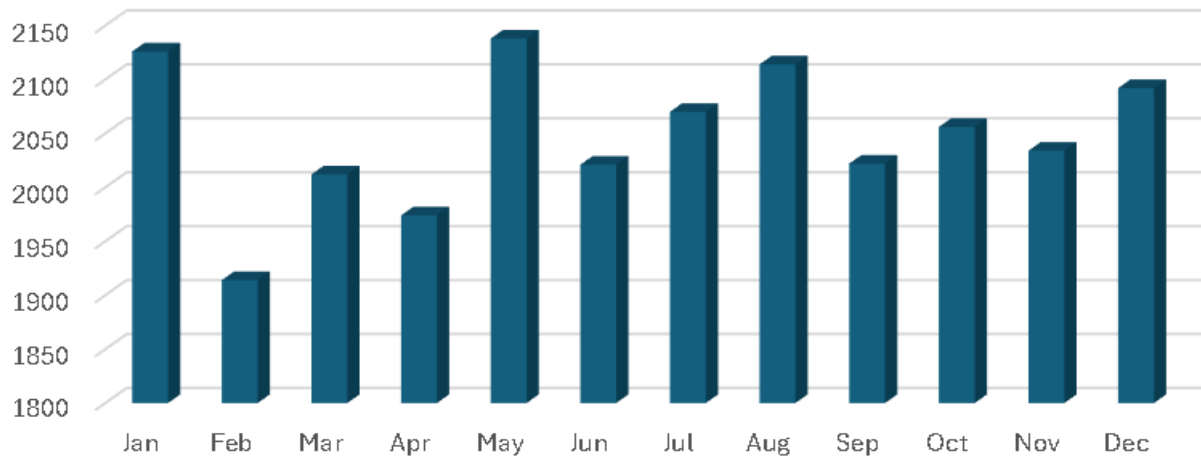


Key Takeaways:

Animals is the top performer, followed by **Science** and **Healthy Eating**. **Technology** and **Food** round out the top 5, all showing strong engagement.

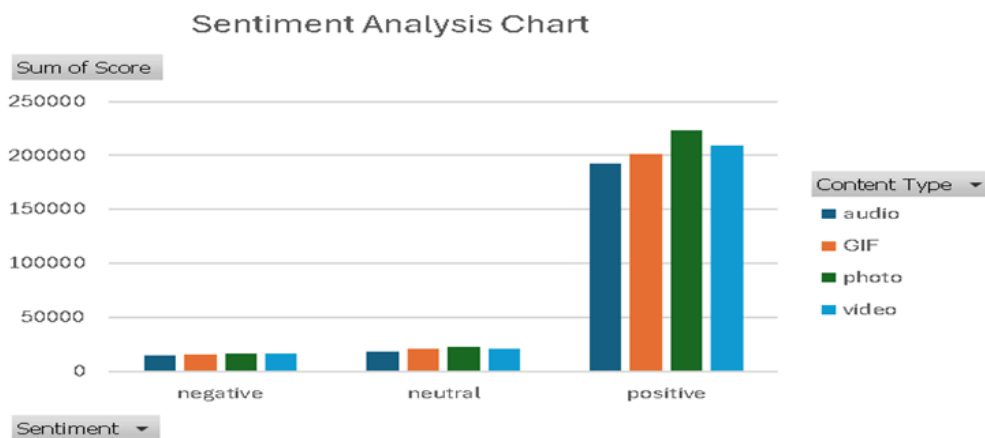
Let's count the monthly content and use a bar chart to represent the “monthly performance”

Monthly Content Performance



- **Peak Activity:** January and May saw the highest levels of activity.
- **Dip in February:** There was a noticeable decrease in activity during February.
- **Stable Months:** Engagement remained consistent around the same level for most months, from June to December.
- **General Trend:** A slight decline in the early year, with steady engagement throughout the rest of the year.

Lastly, we will look at the sentiments each content type received.



User interaction with videos is strong.

Summary

The analysis of Social Buzz's content performance reveals that Animals and Science are the top two categories driving the highest engagement, suggesting a strong interest in educational and lifestyle content. Healthy Eating, Technology, and Food also rank highly, but with slightly less consistent engagement. The overall trend shows stable content activity year-round, with peaks in January and May and a slight dip in February. Furthermore, users express more sentiment reactions to videos contents. These findings highlight key areas for growth and provide actionable insights for refining the company's content strategy moving forward.

Share





Today's agenda

Project recap
Problem
The Analytics team
Process
Insights
Summary



Project Recap

Social Buzz is a fast-growing technology company that needs to manage its current scale. Accenture has begun a 3-month initial project focusing on these tasks:

- An audit of Social Buzz's big data practice
- Recommendations for a successful IPO
- Analysis to find Social Buzz's top 5 popular categories of content

Problem

Over 100,000 pieces of content

36,5000,000 Content per year

How to capitalize on it when there is so much?

Analysis to find Social Buzz's top 5 popular categories of content



The Analytics team



Andrew Fleming
(Chief Technical Architect)



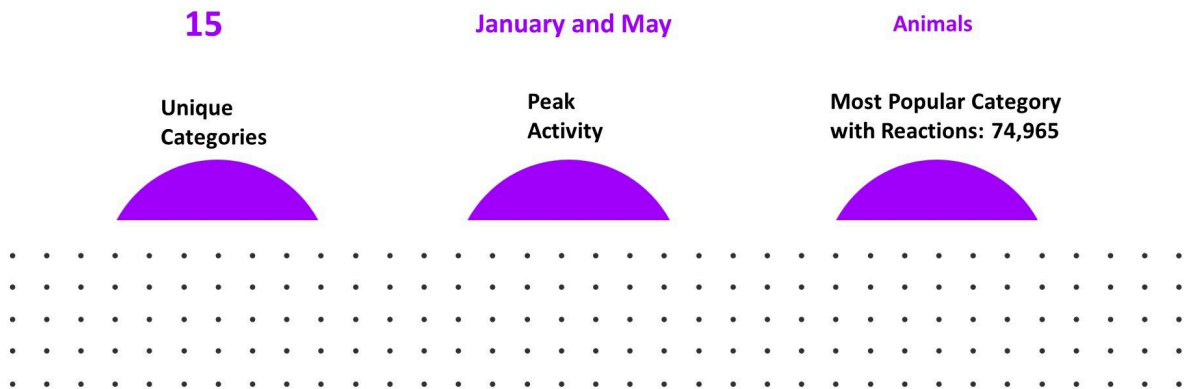
Marcus Rompton
(Senior Principle)



Nelly Moja (Data Analyst)

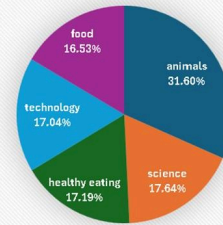


Insights



- **Key Takeaways:**
- **Animals** is the top performer, followed by **Science** and **Healthy Eating**.
- **Technology** and **Food** round out the top 5, all showing strong engagement.

Percentage Share of the Top 5 performing categories



- **Peak Activity:** January and May saw the highest levels of activity.
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- **Stable Months:** Engagement remained consistent around the same level for most months, from June to December.
- **General Trend:** A slight decline in early year, with steady engagement throughout the rest of the year.

Monthly Content Performance



Summary



Recap of Findings:

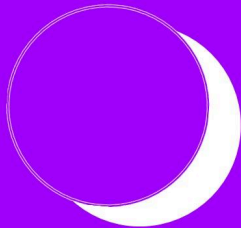
- Animals and Science are the top two performers, and the engagement suggests a growing interest in both educational and lifestyle content.

Actionable Insights:

- Focus content strategy on categories like Animals and Science, as they have the highest popularity.
- Consider further engagement strategies for other categories like Technology and Healthy Eating.

Example: post 30 second videos about healthy eating or technology related topics.

- Conduct another content analysis at a later phase to check on the success of this plan.



Thank you!

ANY QUESTIONS?