Twitter Text Pattern Report

The task of this report and machine learning algorithms is to find the most popular implicit topics in the twtters toward ChatGpt from January 2023 to March 2023, the pattern of the sentiments of people toward ChatGpt, How it differs across three months of 2023, the conversations around ChatGpt on Twitter and how it has changed over time.

The methods of machine learning algorithms implemented in this report include sentiment analysis, the unsupervised models Lda and Word2vec, and the semi-supervised model Corex based on the multiple tasks of this report and desired outcome of the algorithm.

1.Text Data

The source of the dataset comes from Kaggle, and it contains a csv file related to ChatGpt, including the keyword chatgpt,#hashtag and mentions about ChatGpt. The author of this dataset is Khalid AnsariA, who is a scholar from New York University and the Collaborators is Khalid Ansari. The number of tweets collected in the dataset includes information on 500,000 tweets. The timeframe started from January 4, 2023 to March 29th 2023. The unique accounts are tweets per uses, snscrape package was implemented during the process of the data collection and no null values identified. The initial dataset comprises 6 columns, "id", "date", "content", "like_count", "retweet_count" and 500031 rows. The datasets aim to determine Al-powered conversational technologies by analysing tweet volume, sentiments, and user engagement.

2.Data Pre-Processing

Six kinds of data cleaning methods were used, including punctuation, tokenization, lowercase, stopword removal, limmatization, and stemming. More specifically, I removed punctuation by using the remove_punct function with library re, removed URLS and emojis in tokenization using library re, removed stopwords using nltk, changed words to lowercase using lower function, Limmatization of Tweet using ntlk.WordNetLemmatizer (), removal of suffices, such as "ing","ly","s", etc using PorterStemmer from the library NLTK for stemming.

3.Method

3.1 Sentiment Analysis

Converted the "date "column to datetime format, resampled the data by day and compute the sentiment count, and plot the data using seaborn after resetting the index of the data frame and melt the sentiment count. Using SentimentIntensityAnalyzer to replace bigrams indicating ChatGpt Positive as cpos and ChatGpt negative as cneg, assigned polarity score as cpos = -3 and cneg = 3 and Defined sentiments based on intensity score, shown as Figure 2, showing overly positive: >0.75, Positive: between 0.05 &0.75, Overly negative: <0.75, Negative: between -0.05 && -0.75, Neutral: 0.05 to 0.05. In order to generate funnel chart of sentiment distribution (Figure 3&4 in Appendix), Setting feature of dataset "content_lemmatized" as value, "Wordcloud" (shown in Figure 5) was used to quickly identified the most important themes in the large body of text.

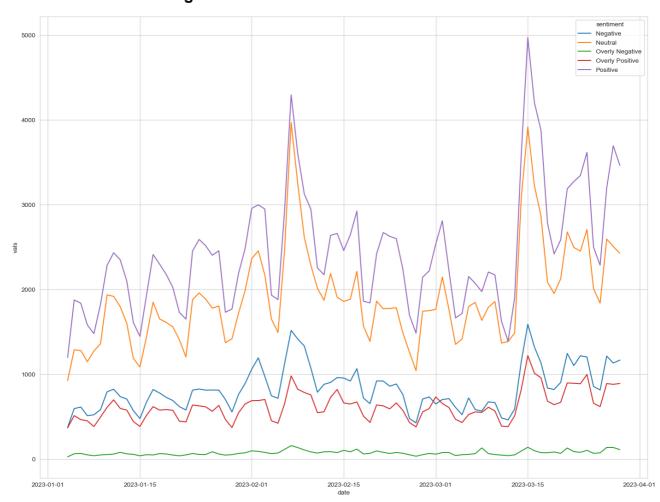
3.2 Unsupervised: Lda and Word2Vec in genism

Built Lda to extract top topics from the text, and computed a probability score for how likely a tweet belongs to each other's topic. Computed the total number of worked and unique keywords. In order to get most similar words related to "chatgpt", Word2Vec and keyedvectors from the genism model was used.

3.3 Semi-Supervised: CorEx

Transformed data into a sparse matrix using Countvectorizer. Set the number of topics as in hidden latent topics as 14 and feature: anchor_strength (how much weight CorEx puts towards maximizing mutual information between anchor words and their respective topics). Normalized topic correlations within individual documents explained by a particular topic. In order to generate normalized topic correlations to represent the correlations within an individual document explained by a particular topic (shown in Figure 6 in the Appendix).

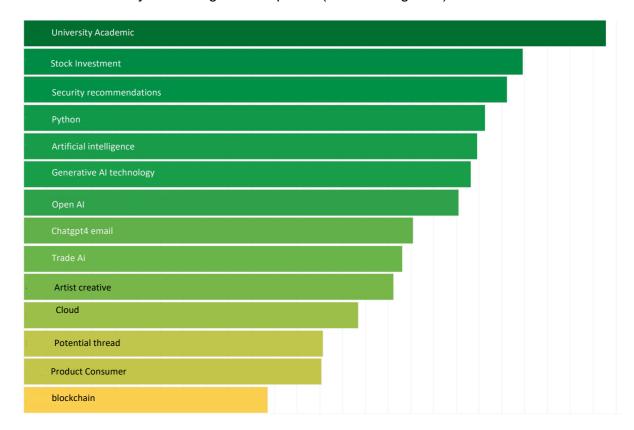
4. Evaluation and findings



Fingure1: Sentiment Analyse from Jan.1 2023 to April.1 2023

February had the lowest number of tweets as compared to January and March in the dataset. 7th February 2023 and 15th March had the highest number of tweets as compared to 26th January and 04th January(lowest) based on the information from Figure 8-9). The pattern of sentiment analysis from 04 January 2023 to 29th March 2023 showed fluctuations between "Neutral", "Negative",

"Overly Positive", and "Positive", with a on 7th Feb 2023. However, the category of "Overly Negative "remained consistently low throughout this period (shown in Figure 1).



Fingure2: Most popular topics between Jan 1 2023 to April 1 2023

Topics identified include "university academic", "stock investment", "security recommendations", "python", "artificial intelligence", "generative ai technology", "open AI", "chatgpt 4 email", "Trade AI", "Artist creative", "Potential Thread", "Cloud", "Product consumer", "Blockchain", according to the Topic Correlation Table (Shown in Figure 6 and Figure 2).

The most related words to "ChatGpt" were "ai", "open ai", "chatgpts", "software", "brickwall", "chatbot", "technology" from word2vec model, and according to WorldCloud Image(Figure 5), the larger the word in the visual, the more common the word was in the tweet text, "chatgpt", "open ai", "artificial intelligence" appear frequently.

The Pearson correlation coefficient between likes and retweets is 0.73, indicating that there is a significant positive relationship between likes and retweets based on the regression plot(Figure 7 in the Appendix).

Conclusion

Results obtained are based on collected data over three months. Positive sentiment contributed the most in overall sentiment(208066), followed by neutral(162388) and negative(70386) sentiments. People were discussing most about topics such as: utilized in university academic, stock investment and security recommendation, Topics that remained underrepresented include artist creative and Cloud. These findings can help understand specific topics and human sentiments creating greater traction ,and how ChatGpt could be utilized in artistic creativity and advance Cloud could be taken into higher consideration by developers and stakeholders.

Appendix

	sentiment	content_lemmatized	;		like_count	retweet_count	score
4	Positive	208066		sentiment			
1	Neutral	162388		Negative	6.733516	1.185119	-0.376120
0	Negative	70387		Neutral	5.543252	1.019565	0.000151
3	Overly Positive	52853		Overly Negative	4.888363	0.972564	-0.827867
	•	Overly Negative 6342		Overly Positive	7.651978	1.816832	0.837100
2	Overly Negative			Positive	8.421495	1.872544	0.446248

Figure 2: Sentiment Score



Figure 3:Funnel-Cahrt of Sentiment Distribution

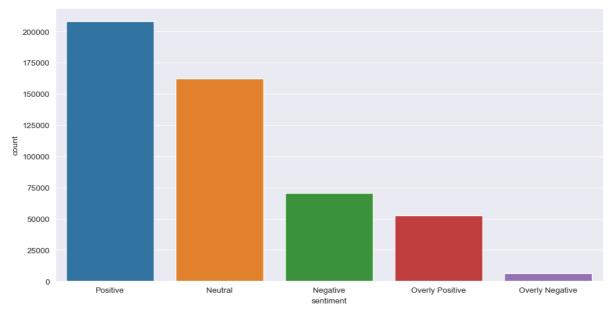


Figure 4: Sentiment Diagram

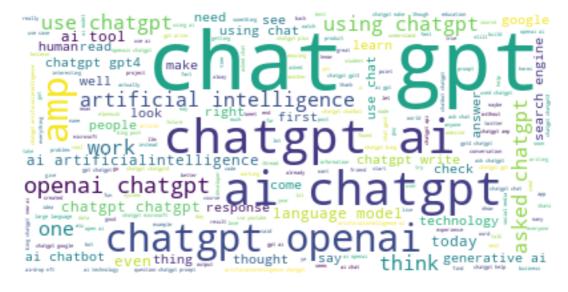


Figure 5: WorldCloud Image

```
0: tool, amp, learning, student, education, school, teacher, essay, edtech, use
1: human, data, look, news, text, research, stock, test, investment, exam
2: asked, answer, question, im, good, ask, code, response, got, problem
3: new, artificialintelligence, google, microsoft, tech, bing, search, bard, company, app
4: tech, based, generated, current, speed, china, instruction, pm, browser, edge
5: use, using, tool, make, way, people, help, used, data, year
6: read, check, article, artificial, midjourney, generativeai, story, generated, blog, written
7: artificialintelligence, tool, model, ai, language, content, gpt3, machinelearning, chatgpt, text
8: crypto, nft, web3, bitcoin, airdrop, blockchain, btc, eth, token, invest
9: artificialintelligence, tech, machinelearning, web3, innovation, python, coding, ml, programming metaverse
10: microsoft, bing, search, free, based, engine, stock, trading, option, buy
11: like, new, time, im, world, thing, day, let, going, great
12: like, make, know, think, work, people, need, human, thing, say
13: new, gpt4, microsoft, model, business, gpt3, version, service, access, api
```

Figure 6: Topic Correlation

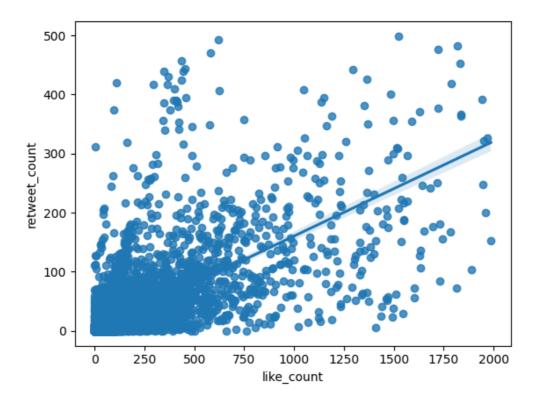


Figure 7: Regression plot

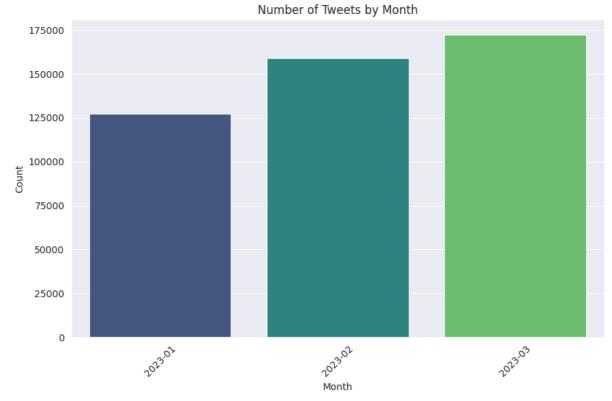


Figure 8: Number of Tweets by Month

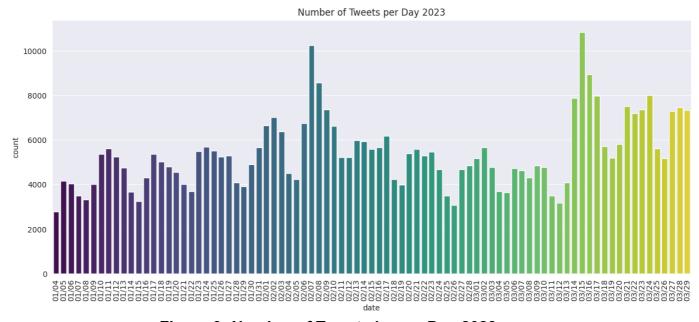


Figure 9: Number of Tweets by per Day 2023

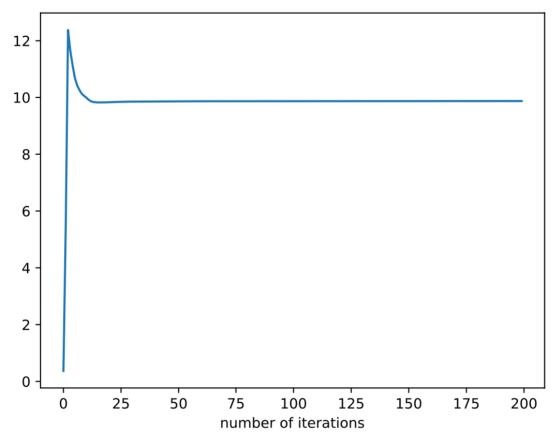


Figure 10: Number of iterations

References

- Portugal, I., Alencar, P. S. C., & Cowan, D. D. (2018). The use of machine learning algorithms in recommender systems: A systematic review. *Expert Systems With Applications*, 97, 205–227. https://doi.org/10.1016/j.eswa.2017.12.020
- Ayodele, T. (2010). Types of Machine Learning Algorithms. In InTech eBooks. https://doi.org/10.5772/9385
- Scaria, V., Patel, H., Nagalapatti, L., Gupta, N., Mehta, S., Guttula, S. C., Mujumdar, S., Afzal, S., Mittal, R. S., & Munigala, V. (2020). Overview and Importance of Data Quality for Machine Learning Tasks. https://doi.org/10.1145/3394486.3406477
- Neethu, M. S., & Rajasree, R. (2013). Sentiment analysis in twitter using machine learning techniques. In 2013 Fourth International Conference on Computing, Communications and Networking Technologies (ICCCNT). https://doi.org/10.1109/icccnt.2013.6726818

- Agarwal, B., & Mittal, N. (2016). Machine Learning Approach for Sentiment Analysis. In Socio-affective computing (pp. 21–45). Springer International

 Publishing. https://doi.org/10.1007/978-3-319-25343-5 3
- Hasan, A. N., Moin, S., Karim, A., & Shamshirband, S. (2018). Machine Learning-Based Sentiment Analysis for Twitter Accounts. Mathematical and Computational Applications, 23(1), 11. https://doi.org/10.3390/mca23010011
- Bojanowski, P., Grave, E., Joulin, A., & Mikolov, T. (2017). Enriching Word Vectors with Subword Information. Transactions of the Association for Computational Linguistics, 5, 135–146. https://doi.org/10.1162/tacl a 00051
- Cui, W., Wu, Y., Liu, S., Wei, F., Zhou, M. X., & Qu, H. (2010). Context preserving dynamic word cloud visualization. https://doi.org/10.1109/pacificvis.2010.5429600
- Heimerl, F., Lohmann, S., Lange, S., & Ertl, T. (2014). Word Cloud Explorer: Text Analytics Based on Word Clouds. https://doi.org/10.1109/hicss.2014.231
- Kumar, P. S., Gupta, D. B., Naha, T. K., & Gupta, S. (2006). Factors affecting fuel rate in Corex process. Ironmaking & Steelmaking, 33(4), 293–298. https://doi.org/10.1179/174328106x101493

Codes From Jupyter

In [2]: import pandas as pd import numpy as np import re import seaborn as sns import matplotlib.pyplot as plt import pandas as pd import string import sys,csv,re from nltk.corpus import stopwords stop_words = set(stopwords.words('english')) from wordcloud import WordCloud from sklearn.feature extraction.text import CountVectorizer from sklearn.model selection import train test split from sklearn.linear model import LogisticRegression from sklearn.metrics import accuracy_score, classification_report, confusion_matrix, ConfusionMatrixDisplay In [3]: df = pd.read csv('/Users/lugiansong/Desktop/TwitterJanMar.csv') In [4]: df.head()

Out[4]:

	date	id	content	username	like_count	retweet_count
0	2023-03-29 22:58:21+00: 00	1641213230730051 584	Free AI marketing and automation tools, strate	RealProfitPros	0.0	0.0
1	2023-03-29 22:58:18+00: 00	1641213218520481 805	@MecoleHardman 4 Chat GPT says it's 15. ⊜	AmyLouWho3 21	0.0	0.0
2	2023-03-29 22:57:53+00: 00	1641213115684536 323	https://t.co/FjJSprt0t e - Chat with any PDF!\n	yjleon1976	0.0	0.0
3	2023-03-29 22:57:52+00: 00	1641213110915571 715	Al muses: "In the court of life, we must all f	ChatGPT_Thi	0.0	0.0
4	2023-03-29 22:57:26+00: 00	1641213003260633 088	Most people haven't heard of Chat GPT yet.\nFi	nikocosmonau t	0.0	0.0

In [5]:

#getting basic information about datasets

df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 500036 entries, 0 to 500035

Data columns (total 6 columns):

Column Non-Null Count Dtype

--- ----- ----

0 date 500036 non-null object
1 id 500030 non-null object
2 content 500030 non-null object
3 username 500002 non-null object
4 like_count 499974 non-null float64
5 retweet_count 499974 non-null float64

dtypes: float64(2), object(4) memory usage: 22.9+ MB

In [6]: df.shape Out[6]: (500036, 6) In [7]:

df.isnull().sum()

Out[7]:

date 0
id 6
content 6
username 34
like_count 62
retweet_count 62

dtype: int64 In [8]: df.columns Out[8]:

Index(['date', 'id', 'content', 'username', 'like_count', 'retweet_count'], dtype='object')

In [9]:

df.reset_index(drop=True)

Out[91:

ԾնելԾյ.	adol.								
	date	id	content	username	like_cou nt	retweet_cou nt			
0	2023-03-29 22:58:21+00: 00	84	Free AI marketing and automation tools, strate	RealProfitPros	0.0	0.0			

	date	id	content	username	like_cou nt	retweet_cou
1	2023-03-29 22:58:18+00: 00	16412132185204818 05	@MecoleHardman4 Chat GPT says it's 15. €	AmyLouWho32	0.0	0.0
2	2023-03-29 22:57:53+00: 00	16412131156845363 23	https://t.co/FjJSprt0t e - Chat with any PDF!\n	e - Chat with any yjleon1976 PDF!\n		0.0
3	2023-03-29 22:57:52+00: 00	16412131109155717 15	Al muses: "In the court of life, we must all f	urt of life, we must		0.0
4	2023-03-29 22:57:26+00: 00	16412130032606330 88	Most people haven't heard of Chat GPT yet.\nFi		0.0	0.0
50003 1	2023-01-04 07:18:08+00: 00	16105360380947578 88	@GoogleAI #LAMDA Versus @OpenAI #ChatGPT ?! Wh	Pup_In_Cup	1.0	0.0
50003 2	2023-01-04 07:17:50+00: 00	16105359616701726 74	#ChatGPT \n\nSo much #Censorship.\n\nNe ver tru	TryingToOffen d	2.0	0.0
50003 3	2023-01-04 07:17:20+00: 00	16105358373634867 20	all my twitter feed is about ChatGPT and @Open	Il my twitter feed is bout ChatGPT and mcp350		1.0
50003 4	2023-01-04 07:17:08+00: 00	16105357860170915 84	I'm quite amazed by Chat GPT. A really promisi manumurali369		1.0	0.0
50003 5	2023-01-04 07:16:56+00: 00	16105357347582197 78	I used chat gpt to get gym workout program and	pnik91	0.0	0.0

500036 rows × 6 columns

In [11]: #All LowerCase

def converter(x):

try:

return ' '.join([x.lower() for x in str(x).split() if x not in stop_words]) except AttributeError:

return None

df['content'] = df['content'].apply(converter)

In [12]: df.head() Out[12]:

	date	id	content	username	like_coun t	retweet_coun t
0	2023-03-29 22:58:21+00:0	164121323073005158 4	free ai marketing automation tools, strategies	RealProfitPros	0.0	0.0
1	2023-03-29 22:58:18+00:0	164121321852048180 5	@mecolehardman 4 chat gpt says it's 15. ⊌	AmyLouWho321	0.0	0.0
2	2023-03-29 22:57:53+00:0	164121311568453632 3	https://t.co/fjjsprt0t e - chat pdf! check new	yjleon1976	0.0	0.0

	date	id	content	username	like_coun t	retweet_coun t
3	2023-03-29 22:57:52+00:0	164121311091557171 5	ai muses: "in court life, must face judge dest	ChatGPT_Think s	0.0	0.0
4	2023-03-29 22:57:26+00:0 0	164121300326063308 8	people heard chat gpt yet. first, elite factio	nikocosmonaut	0.0	0.0

In [13]:

#Removing Punctuation

df['content_punc'] =df['content'].str.replace('[^\w\s]',")

df.head()

/var/folders/m_/m3lsq_594494n7k5zm6nmdtc0000gn/T/ipykernel_40873/787010310.py:2: FutureWarning:

The default value of regex will change from True to False in a future version.

df['content_punc'] =df['content'].str.replace('[^\w\s]',")

Out[13]:

	date	id	content	username	like_cou nt	retweet_co unt	content_punc
0	2023-03-29 22:58:21+00 :00	164121323073005 1584	free ai marketing automation tools, strategies	RealProfitPro s	0.0	0.0	free ai marketing automation tools strategies
1	2023-03-29 22:58:18+00 :00	164121321852048 1805	@mecolehardm an4 chat gpt says it's 15.	AmyLouWho3 21	0.0	0.0	mecolehardm an4 chat gpt says its 15
2	2023-03-29 22:57:53+00 :00	164121311568453 6323	https://t.co/fjjsprt 0te - chat pdf! check new	yjleon1976	0.0	0.0	httpstcofjjsprt0 te chat pdf check new ai quic
3	2023-03-29 22:57:52+00 :00	164121311091557 1715	ai muses: "in court life, must face judge dest	ChatGPT_Thi	0.0	0.0	ai muses in court life must face judge destiny
4	2023-03-29 22:57:26+00 :00	164121300326063 3088	people heard chat gpt yet. first, elite factio	nikocosmona ut	0.0	0.0	people heard chat gpt yet first elite factions

In [14]:

#Removal of stop words

import nltk

from nltk.corpus import stopwords

nltk.download('stopwords')

stop = stopwords.words('english')

df['content_stop'] = df['content_punc'] .apply(lambda x: " ".join(x for x in x.split() if x not in stop)) df.head()

[nltk_data] Downloading package stopwords to

[nltk_data] /Users/luqiansong/nltk_data...

[nltk_data] Package stopwords is already up-to-date!

Out[14]:

	1								
	date	id	content	username	like_co unt	retweet_c ount	content_pun	content_sto	
0	2023-03- 29 22:58:21+0 0:00	1641213230730 051584	free ai marketing automation tools, strategies	RealProfitPr os	0.0	0.0	marketing automation tools	free ai marketing automation tools strategies	

	date	id	content	username	like_co unt	retweet_c ount	content_pun c	content_sto
1	2023-03- 29 22:58:18+0 0:00	1641213218520 481805	@mecolehard man4 chat gpt says it's 15.	AmyLouWh o321	0.0	0.0	mecolehard man4 chat gpt says its 15	mecolehard man4 chat gpt says 15
2	2023-03-	1641213115684 536323	https://t.co/fjjs prt0te - chat pdf! check new	yjleon1976	0.0	0.0	httpstcofjjsp rt0te chat pdf check new ai quic	httpstcofjjsp rt0te chat pdf check new ai quick
3	2023-03- 29 22:57:52+0 0:00	1641213110915 571715	ai muses: "in court life, must face judge dest	ChatGPT_T hinks	0.0	0.0	ai muses in court life must face judge destiny	ai muses court life must face judge destiny ju
4	2023-03- 29	1641213003260 633088	people heard chat gpt yet. first, elite factio	nikocosmon aut	0.0	0.0	people heard chat gpt yet first elite factions	people heard chat gpt yet first elite factions

In [15]:

#Tokenization of Tweets

import textblob

from textblob import TextBlob

def tokenization(content):

content = re.split('\W+', content)

return content

df['content_tokenized'] = df['content_stop'].apply(lambda x: tokenization(x.lower()))

df[['content', 'content_punc', 'content_stop', 'content_tokenized']][0:9]

Out[15]:

_	uti roj.							
	content	content_punc	content_stop	content_tokenized				
0	free ai marketing automation tools, strategies	free ai marketing automation tools strategies	free ai marketing automation tools strategies	[free, ai, marketing, automation, tools, strat				
1	@mecolehardman4 chat gpt says it's 15.	mecolehardman4 chat gpt says its 15	mecolehardman4 chat gpt says 15	[mecolehardman4, chat, gpt, says, 15]				
2	https://t.co/fjjsprt0te - chat pdf! check new	httpstcofjjsprt0te chat pdf check new ai quic	httpstcofjjsprt0te chat pdf check new ai quick	[httpstcofjjsprt0te, chat, pdf, check, new, ai				
3	ai muses: "in court life, must face judge dest	ai muses in court life must face judge destiny	ai muses court life must face judge destiny ju	[ai, muses, court, life, must, face, judge, de				
4	people heard chat gpt yet. first, elite factio	people heard chat gpt yet first elite factions	people heard chat gpt yet first elite factions	[people, heard, chat, gpt, yet, first, elite,				
5	@nytimes no! chat gpt putting together amazing	nytimes no chat gpt putting together amazing r	nytimes chat gpt putting together amazing recipes	[nytimes, chat, gpt, putting, together, amazin				
6	@ylzkrtt yes also chat gpt make generative art	ylzkrtt yes also chat gpt make generative art	ylzkrtt yes also chat gpt make generative art	[ylzkrtt, yes, also, chat, gpt, make, generati				
7	@robinhanson @razibkhan people heard chat gpt	robinhanson razibkhan people heard chat gpt ye	robinhanson razibkhan people heard chat gpt ye	[robinhanson, razibkhan, people, heard, chat,				
8	robotically - shaun usher - letters note thi	robotically shaun usher letters note think	robotically shaun usher letters note think cha	[robotically, shaun, usher, letters, note, thi				
In	[16]-							

In [16]:

#Lemmatization is a more effective option than stemming because it converts the word into its root word,

#rather than just stripping the suffices.

#nltk.download('wordnet')

wn = nltk.WordNetLemmatizer()

def lemmatizer(content):

content = [wn.lemmatize(word) for word in content]

return content

df['content_lemmatized'] = df['content_tokenized'].apply(lambda x: lemmatizer(x))

df[['content', 'content_punc', 'content_tokenized','content_stop','content_lemmatized']][0:9]

Out[16]:

	content	content_punc	content_tokenized	content_stop	content_lemmatized
0	free ai marketing automation tools, strategies	free ai marketing automation tools strategies	[free, ai, marketing, automation, tools, strat	free ai marketing automation tools strategies	[free, ai, marketing, automation, tool, strate
1	@mecolehardman4 chat gpt says it's 15.	mecolehardman4 chat gpt says its 15	[mecolehardman4, chat, gpt, says, 15]	mecolehardman4 chat gpt says 15	[mecolehardman4, chat, gpt, say, 15]
2	https://t.co/fjjsprt0te - chat pdf! check new	httpstcofjjsprt0te chat pdf check new ai quic	[httpstcofjjsprt0te, chat, pdf, check, new, ai	httpstcofjjsprt0te chat pdf check new ai quick	[httpstcofjjsprt0te, chat, pdf, check, new, ai
3	ai muses: "in court life, must face judge dest	ai muses in court life must face judge destiny	[ai, muses, court, life, must, face, judge, de	ai muses court life must face judge destiny ju	[ai, mus, court, life, must, face, judge, dest
4	people heard chat gpt yet. first, elite factio	people heard chat gpt yet first elite factions	[people, heard, chat, gpt, yet, first, elite,	people heard chat gpt yet first elite factions	[people, heard, chat, gpt, yet, first, elite,
5	@nytimes no! chat gpt putting together amazing	nytimes no chat gpt putting together amazing r	[nytimes, chat, gpt, putting, together, amazin	nytimes chat gpt putting together amazing recipes	[nytimes, chat, gpt, putting, together, amazin
6	@ylzkrtt yes also chat gpt make generative art	ylzkrtt yes also chat gpt make generative art	[ylzkrtt, yes, also, chat, gpt, make, generati	ylzkrtt yes also chat gpt make generative art	[ylzkrtt, yes, also, chat, gpt, make, generati
7	@robinhanson @razibkhan people heard chat gpt	robinhanson razibkhan people heard chat gpt ye	[robinhanson, razibkhan, people, heard, chat,	robinhanson razibkhan people heard chat gpt ye	[robinhanson, razibkhan, people, heard, chat,
8	robotically - shaun usher - letters note thi	robotically shaun usher letters note think	[robotically, shaun, usher, letters, note, thi	robotically shaun usher letters note think cha	[robotically, shaun, usher, letter, note, thin

In [17]:

df.drop(columns=['content', 'content_punc', 'content_tokenized', 'content_stop'])

Out[17]:

	date	id	username	like_cou nt	retweet_cou nt	content_lemmatiz ed
0	2023-03-29 22:58:21+00:0	16412132307300515 84	RealProfitPros	0.0	0.0	[free, ai, marketing, automation, tool, strate
1	2023-03-29 22:58:18+00:0	16412132185204818 05	AmyLouWho32	0.0	0.0	[mecolehardman4, chat, gpt, say, 15]
2	2023-03-29 22:57:53+00:0	16412131156845363 23	yjleon1976	0.0	0.0	[httpstcofjjsprt0te, chat, pdf, check, new, ai

	date	id	username	like_cou nt	retweet_cou nt	content_lemmatiz ed
3	2023-03-29 22:57:52+00:0	16412131109155717 15	ChatGPT_Thin ks	0.0	0.0	[ai, mus, court, life, must, face, judge, dest
4	2023-03-29 22:57:26+00:0	16412130032606330 88	nikocosmonaut	0.0	0.0	[people, heard, chat, gpt, yet, first, elite,
50003 1	2023-01-04 07:18:08+00:0 0	16105360380947578 88	Pup_In_Cup	1.0	0.0	[googleai, lamda, versus, openai, chatgpt, car
50003 2	2023-01-04 07:17:50+00:0 0	16105359616701726 74	TryingToOffend	2.0	0.0	[chatgpt, much, censorship, never, trust, syst
50003 3	2023-01-04 07:17:20+00:0 0	16105358373634867 20	mcp350	3.0	1.0	[twitter, feed, chatgpt, openai, lol, ai, chat
50003 4	2023-01-04 07:17:08+00:0 0	16105357860170915 84	manumurali369	1.0	0.0	[im, quite, amazed, chat, gpt, really, promisi
50003 5	2023-01-04 07:16:56+00:0 0	16105357347582197 78	pnik91	0.0	0.0	[used, chat, gpt, get, gym, workout, program,

```
500036 rows × 6 columns
```

In [18]:

df.head()

df.to_csv('Tweets_Cleaned.csv')

In [19]:

Import the wordcloud library

from wordcloud import WordCloud

Join the different processed titles together.

long_string = ' '.join(list(df['content_stop'].values))

Create a WordCloud object

wordcloud = WordCloud(background_color="white", max_words=5000, contour_width=3,

contour_color='steelblue')

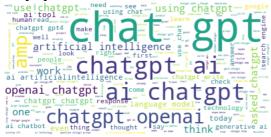
Generate a word cloud

wordcloud.generate(long_string)

Visualize the word cloud

wordcloud.to_image()

Out[19]:



In [20]:

#Stemming refers to the removal of suffices, like "ing", "ly", "s", etc. by a simple rule-based approach. #For this purpose, we will use PorterStemmer from the NLTK library.

from nltk.stem import PorterStemmer

ps = PorterStemmer()

def stemming(content):

content = [ps.stem(word) for word in content]

return content

df['content_stemmed'] = df['content_tokenized'].apply(lambda x: stemming(x))

df[['content', 'content_punc', 'content_tokenized','content_stop','content_stemmed']][0:9] Out[20]:

	content	content_punc	content_tokenized	content_stop	content_stemmed
0	free ai marketing automation tools, strategies	free ai marketing automation tools strategies	[free, ai, marketing, automation, tools, strat	free ai marketing automation tools strategies	[free, ai, market, autom, tool, strategi, coll
1	@mecolehardman4 chat gpt says it's 15.	mecolehardman4 chat gpt says its 15	[mecolehardman4, chat, gpt, says, 15]	mecolehardman4 chat gpt says 15	[mecolehardman4, chat, gpt, say, 15]
2	https://t.co/fjjsprt0te - chat pdf! check new	httpstcofjjsprt0te chat pdf check new ai quic	[httpstcofjjsprt0te, chat, pdf, check, new, ai	httpstcofjjsprt0te chat pdf check new ai quick	[httpstcofjjsprt0t, chat, pdf, check, new, ai,
3	ai muses: "in court life, must face judge dest	ai muses in court life must face judge destiny	[ai, muses, court, life, must, face, judge, de	ai muses court life must face judge destiny ju	[ai, muse, court, life, must, face, judg, dest
4	people heard chat gpt yet. first, elite factio	people heard chat gpt yet first elite factions	[people, heard, chat, gpt, yet, first, elite,	people heard chat gpt yet first elite factions	[peopl, heard, chat, gpt, yet, first, elit, fa
5	@nytimes no! chat gpt putting together amazing	nytimes no chat gpt putting together amazing r	[nytimes, chat, gpt, putting, together, amazin	nytimes chat gpt putting together amazing recipes	[nytim, chat, gpt, put, togeth, amaz, recip]
6	@ylzkrtt yes also chat gpt make generative art	ylzkrtt yes also chat gpt make generative art	[ylzkrtt, yes, also, chat, gpt, make, generati	ylzkrtt yes also chat gpt make generative art	[ylzkrtt, ye, also, chat, gpt, make, gener, ar
7	@robinhanson @razibkhan people heard chat gpt	robinhanson razibkhan people heard chat gpt ye	[robinhanson, razibkhan, people, heard, chat,	robinhanson razibkhan people heard chat gpt ye	[robinhanson, razibkhan, peopl, heard, chat, g
8	robotically - shaun usher - letters note thi	robotically shaun usher letters note think	[robotically, shaun, usher, letters, note, thi	robotically shaun usher letters note think cha	[robot, shaun, usher, letter, note, think, cha

In [47]:

from gensim.models import Word2Vec,KeyedVectors

In [49]:

model=Word2Vec(df.content_lemmatized,min_count=1,vector_size=32)

In [53]:

model.wv.most_similar("chatgpt")

Out[53]:

[('ai', 0.8875566720962524),

('openai', 0.801838219165802),

('chatgpts', 0.7576317191123962),

('here', 0.7523341774940491),

('software', 0.7216390371322632),

('chatgpt4', 0.709638774394989),

('brickwall', 0.7091189026832581),

('chatbot', 0.7029999494552612),

('openais', 0.6793273687362671),

('technology', 0.6696925759315491)]

In [90]:

import numpy as np # linear algebra

import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)

import re

import string

import nltk

import matplotlib.pyplot as plt

import seaborn as sns

sns.set_style('darkgrid')

```
nltk.download('vader lexicon')
from nltk.sentiment.vader import SentimentIntensityAnalyzer as SIA
from wordcloud import WordCloud, STOPWORDS
plt.rc('figure',figsize=(17,13))
import plotly.express as px
import plotly graph objs as go
import plotly.offline as pyo
from plotly.subplots import make subplots
[nltk data] Downloading package vader lexicon to
[nltk data] /Users/lugiansong/nltk data...
[nltk data] Package vader lexicon is already up-to-date!
In [92]:
df['content_lemmatized'] = df['content'].str.replace('[^\w\s]',")
/var/folders/m /m3lsq 594494n7k5zm6nmdtc0000gn/T/ipykernel 40873/2903736268.py:1: FutureWarning:
The default value of regex will change from True to False in a future version.
In [95]:
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
new words = {
  'cpos': -3.0,
  'cneq': 3.0,
analyser = SentimentIntensityAnalyzer()
analyser.lexicon.update(new words)
scores=[]
for i in range(len(df['content lemmatized'])):
  score = analyser.polarity scores(df['content lemmatized'][i])
  score=score['compound']
  scores.append(score)
sentiment=[]
for i in scores:
  if i > = 0.75:
     sentiment.append('Overly Positive')
  elif (i>=0.05) and (i<0.75):
     sentiment.append('Positive')
  elif i<=(-0.75):
     sentiment.append('Overly Negative')
  elif (i<=-0.05) and (i>-0.75):
     sentiment.append('Negative')
  else:
     sentiment.append('Neutral')
df['sentiment']= pd.Series(np.array(sentiment))
df['score']= pd.Series(np.array(scores))
In [97]:
df.head()
Out[97]:
```

	ավօդ.												
	date	id	content	userna me	_	retwe et_co unt	content _punc	content _stop	content _tokeni zed	content_ lemmati zed	content _stemm ed	sent ime nt	sc or e
(2023- 03-29 22:58: 21+00:		free ai marketin g automati on tools, strategie s		0.0	0.0	ng automa tion	free ai marketi ng automa tion tools strategi es	[free, ai, marketi ng, automat ion, tools, strat	free ai marketin g automati on tools strategie s	[free, ai, market, autom, tool, strategi, coll	Pos itive	0. 51 06
1	2023- 03-29 22:58:	16412132 18520481 805	@mecol ehardm an4 chat	uWho3	0.0	0.0	hardma		hardma	mecoleh ardman4 chat gpt	-	Neu tral	0. 00 00

	date	id	content	userna me	like _co unt	retwe et_co unt	content _punc	content _stop	content _tokeni zed	content_ lemmati zed	content _stemm ed	sent ime nt	sc or e
	18+00: 00		gpt says it's 15.				gpt says its 15	gpt says 15	chat, gpt, says, 15]	says its 15	chat, gpt, say, 15]		
2	2023- 03-29 22:57: 53+00: 00	16412131 15684536 323	https://t. co/fjjsprt 0te - chat pdf! check new	yjleon1 976	0.0	0.0	httpstc ofjjsprt Ote chat pdf check new ai quic	httpstc ofjjsprt Ote chat pdf check new ai quick	[httpstc ofjjsprt0 te, chat, pdf, check, new, ai	httpstcof jjsprt0te chat pdf check new ai quic	[httpstc ofjjsprt0 t, chat, pdf, check, new, ai,	Pos itive	0. 71 84
हा	2023- 03-29 22:57: 52+00: 00	16412131 10915571 715	ai muses: "in court life, must face judge dest	ChatG PT_Thi nks	0.0	0.0	ai muses in court life must face judge destiny 	ai muses court life must face judge destiny ju	[ai, muses, court, life, must, face, judge, de	ai muses in court life must face judge destiny	[ai, muse, court, life, must, face, judg, dest	Neu tral	0. 00 00
	2023- 03-29 22:57: 26+00: 00	16412130 03260633 088		nikocos monaut	0.0	0.0	people heard chat gpt yet first elite faction s	people heard chat gpt yet first elite faction s	[people, heard, chat, gpt, yet, first, elite,	people heard chat gpt yet first elite factions.	[peopl, heard, chat, gpt, yet, first, elit, fa	Neu tral	0. 02 58

In [98]:

df.shape

Out[98]: (500036, 13)

Ìn [100]:

df.groupby(by="sentiment").mean()

Out[100]:

	like_count	retweet_count	score
sentiment			
Negative	6.733516	1.185119	-0.376120
Neutral	5.543252	1.019565	0.000151
Overly Negative	4.888363	0.972564	-0.827867
Overly Positive	7.651978	1.816832	0.837100
Positive	8.421495	1.872544	0.446248

In [101]:

temp =

temp.style.background_gradient(cmap='Purples')

Out[101]:

sentiment content_lemmatized

4 Positive 208066

```
sentiment
                          content_lemmatized
1 Neutral
                          162388
0 Negative
                          70387
3 Overly Positive
                          52853
2 Overly Negative
                          6342
In [113]:
plt.figure(figsize=(12,6))
sns.countplot(x='sentiment',data=df)
fig = go.Figure(go.Funnelarea(
  text =temp.sentiment,
  values = temp.content lemmatized,
  title = {"position": "top center", "text": "Funnel-Chart of Sentiment Distribution"}
  ))
fig.show()
In [176]:
import pandas as pd
import time
import datetime
import calendar
import seaborn as sns
import matplotlib.pyplot as plt
from datetime import datetime
In [207]:
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
new words = {
  'cpos': -3.0,
  'cneg': 3.0,
}
analyser = SentimentIntensityAnalyzer()
analyser.lexicon.update(new_words)
scores=[]
for i in range(len(df['content lemmatized'])):
  score = analyser.polarity_scores(df['content_lemmatized'][i])
  score=score['compound']
  scores.append(score)
sentiment=[]
for i in scores:
  if i > = 0.75:
     sentiment.append('Overly Positive')
  elif (i>=0.05) and (i<0.75):
     sentiment.append('Positive')
  elif i < = (-0.75):
     sentiment.append('Overly Negative')
  elif (i<=-0.05) and (i>-0.75):
     sentiment.append('Negative')
  else:
     sentiment.append('Neutral')
```

df['sentiment']= pd.Series(np.array(sentiment))
df['score']= pd.Series(np.array(scores))
df.head()
Out[207]:

Unn ame d: 0	date	id	content	userna me	like_ cou nt	retwe et_co unt	content _punc	content _stop	content_ tokenize d	content_I emmatiz ed	sent ime nt	sc or e
0	2023- 03-29 22:58: 21+00:	16412132 30730051 584	free ai marketin g automati on tools, strategie s	RealPr ofitPros	0.0	0.0	free ai marketi ng automa tion tools strategi es	free ai marketi ng automa tion tools strategi es	['free', 'ai', 'marketi ng', 'automat ion', 'too	['free', 'ai', 'marketin g', 'automati on', 'too	Posi tive	0.5 10 6
1	2023- 03-29 22:58: 18+00: 00	16412132 18520481 805	@mecol ehardma n4 chat gpt says it's 15.	AmyLo uWho3 21	0.0	0.0	mecole hardma n4 chat gpt says its 15	mecole hardma n4 chat gpt says 15	['mecole hardman 4', 'chat', 'gpt', 'says', '15']	['mecole hardman 4', 'chat', 'gpt', 'say', '15']	Neu tral	0.0 00 0
2	2023- 03-29 22:57: 53+00: 00	16412131 15684536 323	https://t. co/fjjsprt 0te - chat pdf! check new	yjleon1 976	0.0	0.0	httpstco fjjsprt0t e chat pdf check new ai quic	httpstco fjjsprt0t e chat pdf check new ai quick	['httpstco fjjsprt0te' , 'chat', 'pdf', 'check',	['httpstco fjjsprt0te' , 'chat', 'pdf', 'check',	Posi tive	0.7 18 4
3	2023- 03-29 22:57: 52+00: 00	16412131 10915571 715	ai muses: "in court life, must face judge dest	ChatG PT_Thi nks	0.0	0.0	ai muses in court life must face judge destiny.	ai muses court life must face judge destiny ju	['ai', 'muses', 'court', 'life', 'must', 'face	['ai', 'mus', 'court', 'life', 'must', 'face',	Neu tral	0.0 00 0
4	2023- 03-29 22:57: 26+00: 00	16412130 03260633 088	people heard chat gpt yet. first, elite factio	nikocos monaut	0.0	0.0	yet first elite	people heard chat gpt yet first elite factions 	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	Neu tral	0.0 25 8

In [208]: df.date Out[208]:

0 2023-03-29 22:58:21+00:00 1 2023-03-29 22:58:18+00:00

2 2023-03-29 22:57:53+00:00

3 2023-03-29 22:57:52+00:00 4 2023-03-29 22:57:26+00:00

500031 2023-01-04 07:18:08+00:00 500032 2023-01-04 07:17:50+00:00 500033 2023-01-04 07:17:20+00:00 500034 2023-01-04 07:17:08+00:00 500035 2023-01-04 07:16:56+00:00 Name: date, Length: 500036, dtype: object

In [232]:

```
from datetime import datetime
# define a function to convert date format
def convert_date_format(date_str):
  try:
     date obj = datetime.strptime(date str, '%Y-%m-%d %H:%M:%S+00:00')
     return date_obj.strftime('%Y-%m-%d')
  except ValueError:
     return date str
# apply the function to the date column
df['date'] = df['date'].apply(convert_date_format)
In [233]:
df.date
Out[233]:
       2023-03-29
0
1
       2023-03-29
2
       2023-03-29
3
       2023-03-29
       2023-03-29
4
500031 2023-01-04
500032 2023-01-04
500033 2023-01-04
500034 2023-01-04
500035 2023-01-04
Name: date, Length: 500036, dtype: object
In [234]:
df.head()
Out[234]:
```

	Unn ame d: 0	da	id	content	userna me	_	retwee t_coun t	content _punc	content _stop	content_t okenized	content_l emmatize d	sent ime nt	sc ore
C	0	20 23 - 03 - 29	164121323 073005158 4	free ai marketin g automati on tools, strategie s	RealPro fitPros	0.0	0.0		free ai marketi ng automat ion tools strategi es		['free', 'ai', 'marketin g', 'automati on', 'too	Posi tive	0.5 10 6
1	1	20 23 - 03 - 29	164121321 852048180 5	@mecol ehardma n4 chat gpt says it's 15.	AmyLou Who32 1	0.0	0.0	mecole hardma n4 chat gpt says its 15	mecole hardma n4 chat gpt says 15	['mecole hardman 4', 'chat', 'gpt', 'says', '15']	['mecoleh ardman4' , 'chat', 'gpt', 'say', '15']	Neu tral	0.0 00 0
2	2	20 23 - 03 - 29	164121311 568453632 3	https://t.c o/fjjsprt0t e - chat pdf! check new	yjleon1 976	0.0	0.0	httpstco fjjsprt0t e chat pdf check new ai quic	httpstco fjjsprt0t e chat pdf check new ai quick	['httpstco fjjsprt0te' , 'chat', 'pdf', 'check',	['httpstcof jjsprt0te', 'chat', 'pdf', 'check',	Posi tive	0.7 18 4
3	3	20 23 - 03 - 29	164121311 091557171 5	ai muses: "in court life, must face judge dest	ChatGP T_Think s	0.0	0.0	ai muses in court life must face judge destiny	ai muses court life must face judge destiny ju	['ai', 'muses', 'court', 'life', 'must', 'face	['ai', 'mus', 'court', 'life', 'must', 'face',	Neu tral	0.0 00 0

	ame	da te	id	content	userna me	_	retwee t_coun t	content		content_t okenized	content_l emmatize d	sent ime nt	sc ore
4	4	20 23 - 03 - 29	164121300	people heard chat gpt yet. first, elite factio	nikocos monaut	0.0	0.0	yet first elite	chat gpt yet first elite	'heard', 'chat', 'gpt',	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	Neu tral	0.0 25 8

In [237]:

df = df[pd.to_datetime(df['date'], errors='coerce').notnull()]

convert the date column to datetime format
df['date'] = pd.to_datetime(df['date'])

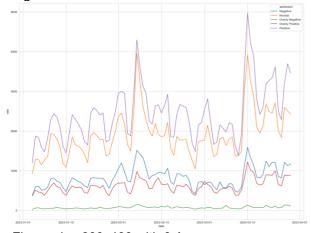
resample the data by day and compute the sentiment count timeline = df.resample('D', on='date')['sentiment'].value_counts().unstack(1)

reset the index of the dataframe and melt the sentiment columns timeline.reset_index(inplace=True) timeline = timeline.melt('date', var_name='sentiment', value_name='vals')

plot the data using seaborn import seaborn as sns import matplotlib.pyplot as plt

sns.set_style('whitegrid')
sns.lineplot(x='date', y='vals', hue='sentiment', data=timeline)
plt.figure(figsize=(6,4))
Out[237]:

<Figure size 600x400 with 0 Axes>



<Figure size 600x400 with 0 Axes>

```
# Static Day
tweets_by_day = df.groupby(pd.Grouper(key='date', freq='D')).size().reset_index()
tweets_by_day.columns = ['date', 'count']
tweets_by_day['date'] = tweets_by_day['date'].dt.strftime('%m/%d')
plt.figure(figsize=(15, 6))
sns.barplot(data=tweets_by_day, x='date', y='count', palette='viridis')
plt.title('Number of Tweets per Day 2023')
plt.xticks(rotation=90)
plt.show()
# Static plot
# Regression plot to understand the relationship between likes and retweets
```

sns.regplot(x='like_count', y='retweet_count', data=df_filtered)
Static Plot
plt.figure(figsize=(12, 6))
sns.histplot(data=polarity_df, bins=40, kde=False, alpha=0.5, palette=['#1DA1F2', '#00CC96'])
plt.title('Distributions of sentimental polarities Vader Vs. TextBlob')
plt.xlabel('Polarity')
plt.ylabel('Count')
plt.show()
In [5]:
import numpy as np
import scipy.sparse as ss
import matplotlib.pyplot as plt

import pandas as pd

import corextopic.corextopic as ct

import corextopic.vis_topic as vt # jupyter notebooks will complain matplotlib is being loaded twice

 $from \ sklearn.feature_extraction.text \ import \ Count Vectorizer$

In [6]:

news data = pd.read csv("/Users/lugiansong/Desktop/newtwitter.csv")

/var/folders/m_/m3lsq_594494n7k5zm6nmdtc0000gn/T/ipykernel_7251/93900312.py:1: DtypeWarning:

Columns (1) have mixed types. Specify dtype option on import or set low memory=False.

news_data = pd.read_csv("/Users/luqiansong/Desktop/newtwitter.csv")

In [7]:

news_data.head()

Out[7]:

	dat e	id	content	username	like_c ount	retweet _count	content_p unc	content_s top	content_to kenized	content_le mmatized
C	20 23- 03- 29	16412132307 30051584	free ai marketing automation tools, strategies	RealProfit Pros	0.0	0.0	free ai marketing automatio n tools strategies 	free ai marketing automatio n tools strategies 	['free', 'ai', 'marketing', 'automation', 'too	['free', 'ai', 'marketing', 'automation ', 'too
1	20 23- 03- 29	16412132185 20481805	@mecoleh ardman4 chat gpt says it's 15.	AmyLou Who321	0.0	0.0	mecoleha rdman4 chat gpt says its 15	mecoleha rdman4 chat gpt says 15	['mecoleha rdman4', 'chat', 'gpt', 'says', '15']	['mecolehar dman4', 'chat', 'gpt', 'say', '15']
2	20 23- 03- 29	16412131156 84536323	https://t.co/ fjjsprt0te - chat pdf! check new	yjleon197 6	0.0	0.0	httpstcofjj sprt0te chat pdf check new ai quic	httpstcofjj sprt0te chat pdf check new ai quick	['httpstcofjj sprt0te', 'chat', 'pdf', 'check',	['httpstcofjjs prt0te', 'chat', 'pdf', 'check',
3	20 23- 03- 29	16412131109 15571715	ai muses: "in court life, must face judge dest	ChatGPT _Thinks	0.0	0.0	ai muses in court life must face judge destiny	ai muses court life must face judge destiny ju	['ai', 'muses', 'court', 'life', 'must', 'face	['ai', 'mus', 'court', 'life', 'must', 'face',
4	20 23- 03- 29	16412130032 60633088	most people heard chat gpt yet. first, elite f	nikocosm onaut	0.0	0.0	most people heard chat gpt yet first elite fac	people heard chat gpt yet first elite factions	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	['people', 'heard', 'chat', 'gpt', 'yet', 'fir

In [8]:

news_data['content_lemmatized1'] = news_data['content_lemmatized'].str.replace('[^\w\s]',")

In [9]:

#Removal of stop words

import nltk

from nltk.corpus import stopwords

nltk.download('stopwords')

stop = stopwords.words('english')

[nltk_data] Downloading package stopwords to

[nltk_data] /Users/luqiansong/nltk_data...

[nltk_data] Package stopwords is already up-to-date!

In [10]:

new = ('12', 'academic', 'teach', 'educator', 'assignment', 'class', 'using', 'use', 'university', 'teaching', 'robot', 'tool', 'edtech', 'essay', 'learning', 'teacher', 'school', 'amp', 'education', 'student', 'billion', 'tool', 'available', 'access', 'business', 'released', 'next', 'update', 'gpt3', 'product', 'plus', 'feature', 'customer', 'model', 'version', 'service', 'api', 'new', 'microsoft', 'gpt4',)

In [11]:

news_data['Tokens'] = news_data['content_lemmatized1'] .apply(lambda x: " ".join(x for x in x.split() if x not in stop))

news_data.head()

Out[11]:

Jui	ъ.	.1.										
d te	la e	id	content	userna me	like_ cou nt	retwe et_co unt	content _punc	content _stop	content_ tokenize d	content_l emmatiz ed	content_l emmatize d1	Tokens
2 - 0 -	20 23 33 29	16412132 30730051 584	free ai marketin g automati on tools, strategie s	RealPr ofitPros	0.0	0.0	free ai marketi ng automa tion tools strategi es	free ai marketi ng automa tion tools strategi es	['free', 'ai', 'marketi ng', 'automat ion', 'too	['free', 'ai', 'marketin g', 'automati on', 'too	['free', 'ai', 'marketin g', 'automati on', 'too	['free', 'ai', 'marketi ng', 'automat ion', 'too
0	20 23 3 29	16412132 18520481 805	@mecol ehardma n4 chat gpt says it's 15.	AmyLo uWho3 21	0.0	0.0	mecole hardma n4 chat gpt says its 15	mecole hardma n4 chat gpt says 15	['mecole hardman 4', 'chat', 'gpt', 'says', '15']	['mecole hardman 4', 'chat', 'gpt', 'say', '15']	['mecoleh ardman4' , 'chat', 'gpt', 'say', '15']	['mecole hardman 4', 'chat', 'gpt', 'say', '15']
2 - 0 -	20 23 3 29	16412131 15684536 323	https://t. co/fjjsprt 0te - chat pdf! check new	yjleon1 976	0.0	0.0		httpstco fjjsprt0t e chat pdf check new ai quick	['httpstc ofjjsprt0t e', 'chat', 'pdf', 'check',	['httpstco fjjsprt0te' , 'chat', 'pdf', 'check',	['httpstcof jjsprt0te', 'chat', 'pdf', 'check',	['httpstc ofjjsprt0t e', 'chat', 'pdf', 'check',
2 - 0 -	20 23 33 29	16412131 10915571 715	ai muses: "in court life, must face judge dest	ChatG PT_Thi nks	0.0	0.0	ai muses in court life must face judge destiny.	ai muses court life must face judge destiny ju	['ai', 'muses', 'court', 'life', 'must', 'face	['ai', 'mus', 'court', 'life', 'must', 'face',	['ai', 'mus', 'court', 'life', 'must', 'face',	['ai', 'mus', 'court', 'life', 'must', 'face',
2 - 0 -	20 23 3 29	16412130 03260633 088	most people heard chat gpt yet. first, elite f	nikocos monaut	0.0	0.0	most people heard chat gpt yet first elite fac	people heard chat gpt yet first elite factions 	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	['people', 'heard', 'chat', 'gpt', 'yet', 'fir	['people', 'heard', 'chat', 'gpt', 'yet', 'fir

In [12]:

[#] Transform data into a sparse matrix

```
vectorizer = CountVectorizer(stop words='english', max features=None, binary=True)
doc word = vectorizer.fit transform(news data.Tokens)
doc word = ss.csr matrix(doc word)
doc_word.shape # n_docs x m_words
Out[12]:
(500036, 619062)
In [13]:
doc word
Out[13]:
<500036x619062 sparse matrix of type '<class 'numpy.int64'>'
         with 6817072 stored elements in Compressed Sparse Row format>
In [16]:
# Get words that label the columns (needed to extract readable topics and make anchoring easier)
words = list(np.asarray(vectorizer.get feature names out()))
In [19]:
anchor_words = [
# academic university help
['12', 'academic', 'teach', 'educator', 'assignment', 'class', 'using', 'use', 'university', 'teaching', 'robot', 'tool',
'edtech', 'essay', 'learning', 'teacher', 'school', 'amp', 'education', 'student'],
# stcok investiment
['stock', 'investment', 'pas', 'state', 'medical', 'look', 'text', 'law', 'algorithm', 'investing', 'news', 'training', 'take',
'trained', 'paper', 'research', 'test', 'human', 'data', 'exam'],
# answer question give response
['problem', 'im', 'would', 'code', 'first', 'try', 'think', 'response', 'asking', 'got', 'give', 'time', 'good', 'one', 'like', 'get',
'ask', 'asked', 'question', 'answer'],
# artificialintelligence company business
['extension', 'impact', 'article', 'new', 'amp', 'read', 'company', 'war', 'googlebard', 'app', 'competitor', 'new',
'rival', 'googleai', 'tech', 'bardai', 'launch', 'apple', 'market', 'artificialintelligence', 'bing', 'microsoft', 'search',
'bard', 'google'],
# instruction of generated ai
['tempupdate', 'kmhr', 'temp', 'wind', 'browser', 'tech', 'chinese', 'baidu', 'status', 'bot', 'edge', 'generated',
'sunrise', 'sunset', 'instruction', 'based', 'speed', 'pm', 'current', 'china'],
# security recommendation
['gonna', '100', 'hour', 'case', 'way', 'used', 'help', 'million', 'much', 'people', 'money', 'year', 'month', 'tool',
'information', 'using', 'time', 'data', 'make', 'user', 'use', 'recommendation', 'date', 'security', 'pro',
'httpstcorlyimpqw40', 'price', 'stablediffusion2', 'powered', 'open', 'wait', 'long', 'cybersecurity', 'gpt', 'last',
'midjourney', 'join', 'dalle', 'short', 'imagine'],
# artists
['digitalart', 'script', 'style', 'used', 'article', 'written', 'poem', 'book', 'story', 'asked', 'write"using', 'another',
'tweet', 'word', 'make', 'fake', 'wrote', 'read', 'writing', 'song', 'generated', 'risk', 'world', 'podcast', 'check', 'blog', 'generativeai', 'tech', 'business', 'artificial', 'bill', 'new', 'aiartwork', 'tesla', 'world', 'woke', 'joke', 'artist', 'image',
'dalle2', 'stablediffusion', 'musk', 'twitter', 'elon', 'aiart', 'art', 'midjourney', 'elonmusk'],
# writing email
['website', 'industry', 'generative', 'response', 'processing', 'machinelearning', 'developed', 'generative',
'chatbots', 'gpt4', 'nlp', 'text', 'natural', 'ability', 'chatgpt3', 'large', 'source', 'domain', 'artificialintelligence', 'llm',
'gpt3', 'model', 'language', 'potential', 'latest', 'future', 'artificialintelligence', 'generate', 'blog', 'email', 'check',
'help', 'amp', 'writing', 'youtube', 'seo', 'via', 'using', 'video', 'create', 'tool', 'use', 'marketing', 'prompt', 'content'],
# future blockchain
['bnb', 'ethereum', 'cryptocurrency', 'powerful', 'nfts', 'invest', 'time', 'magic', 'coin', 'token', 'future', 'eth',
'blockchain', 'bitcoin', 'btc', 'web3', 'airdrop', 'gpt4', 'nft', 'crypto'],
# code python
['new', 'best', 'developer', 'code', 'artificialintelligence', 'programming', 'coding', 'machinelearning', 'learning',
'build', 'free', 'business', 'get', 'datascience', 'tech', 'learn', 'python', 'innovation', 'analytics', 'productivity', 'skill',
'ml', 'latest', 'python', '5g', 'iot', 'say', 'nft', 'lol', 'gaming', 'friend', 'ar', 'dan', 'resume', 'tech', 'vr', 'web3', 'gt',
'artificialintelligence', 'metaverse'],
# trade ai
['altman', 'based', 'bingai', 'nocode', 'sam', 'signal', 'chart', 'trade', 'new', 'trial', 'buy', 'option', 'free', 'stock',
'trading', 'engine', 'microsoft', 'search', 'bing', 'gpt4'],
# potential thread
['come', 'keep', 'new', 'world', 'time', 'take', 'great', 'life', 'one', 'thread', 'thing', 'mind', 'going', 'change', 'like',
'conversation', 'let', 'see', 'im', 'day'],
# cloud job
```

```
['get', 'use', 'work', 'could', 'right', 'make', 'say', 'dont', 'need', 'thing', 'even', 'would', 'replace', 'cant', 'like',
'people', 'human', 'think', 'know', 'job'],
# product consumer
['billion', 'tool', 'available', 'access', 'business', 'released', 'next', 'update', 'gpt3', 'product', 'plus', 'feature',
'customer', 'model', 'version', 'service', 'api', 'new', 'microsoft', 'gpt4']]
anchored topic model = ct.Corex(n hidden=14, seed=2)
anchored topic model.fit(doc word, words=words, anchors=anchor words, anchor strength=2);
WARNING: Anchor word not in word column labels provided to CorEx: take
WARNING: Anchor word not in word column labels provided to CorEx: would
WARNING: Anchor word not in word column labels provided to CorEx: first
WARNING: Anchor word not in word column labels provided to CorEx: give
WARNING: Anchor word not in word column labels provided to CorEx: one
WARNING: Anchor word not in word column labels provided to CorEx: get
WARNING: Anchor word not in word column labels provided to CorEx: much
WARNING: Anchor word not in word column labels provided to CorEx: last
WARNING: Anchor word not in word column labels provided to CorEx: writeusing
WARNING: Anchor word not in word column labels provided to CorEx: another
WARNING: Anchor word not in word column labels provided to CorEx: bill
WARNING: Anchor word not in word column labels provided to CorEx: via
WARNING: Anchor word not in word column labels provided to CorEx: get
WARNING: Anchor word not in word column labels provided to CorEx: keep
WARNING: Anchor word not in word column labels provided to CorEx: take
WARNING: Anchor word not in word column labels provided to CorEx: one
WARNING: Anchor word not in word column labels provided to CorEx: see
WARNING: Anchor word not in word column labels provided to CorEx: get
WARNING: Anchor word not in word column labels provided to CorEx: could
WARNING: Anchor word not in word column labels provided to CorEx: even
WARNING: Anchor word not in word column labels provided to CorEx: would
WARNING: Anchor word not in word column labels provided to CorEx: cant
WARNING: Anchor word not in word column labels provided to CorEx: next
In [26]:
for n in range(len(anchor words)):
  topic_words,__, = zip(*anchored_topic_model.get_topics(topic=n))
  print('{}: '.format(n) + ', '.join(topic_words))
0: tool, amp, learning, student, education, school, teacher, essay, edtech, use
1: human, data, look, news, text, research, stock, test, investment, exam
2: asked, answer, question, im, good, ask, code, response, got, problem
3: new, artificialintelligence, google, microsoft, tech, bing, search, bard, company, app
4: tech. based, generated, current, speed, china, instruction, pm. browser, edge
5: use, using, tool, make, way, people, help, used, data, year
6: read, check, article, artificial, midjourney, generativeai, story, generated, blog, written
7: artificialintelligence, tool, model, ai, language, content, gpt3, machinelearning, chatgpt, text
8: crypto, nft, web3, bitcoin, airdrop, blockchain, btc, eth, token, invest
9: artificialintelligence, tech, machinelearning, web3, innovation, python, coding, ml, programming,
metaverse
10: microsoft, bing, search, free, based, engine, stock, trading, option, buy
11: like, new, time, im, world, thing, day, let, going, great
12: like, make, know, think, work, people, need, human, thing, say
13: new, gpt4, microsoft, model, business, gpt3, version, service, access, api
In [27]:
vt.vis_rep(anchored_topic_model, column_label=words, prefix='twitters')
Print topics in text file
```