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Procedures:

For this project, we collected 3 days' worth of tweets (100 tweets/hour x 24 hours x 3 days x 2 keywords = 14.4K using the model we have built from the last project. The aim of this project is to apply the text mining methods to analyze the topics and popular words that related to the two brand Microsoft and Amazon. In the following report, we will provide detailed procedures and insightful information based on the sentiment analysis and the word cloud we have made.

Before stepping into the requirements, we designed a series of helper functions to set up. This includes functions like removing punctuation and digits, remove stop-words, remove meaningless words("https","co","rt","http",'amp','de','theyr'), loading data from relative path, Stemming and so on. We do this because it will make the upcoming steps a lot easier.

A. [Preliminary Analysis]

Find the ten most popular words with and without stop words

- We did two versions of analysis on the most popular words, one is with stop words, the other is without stop words.
- For the version with stop words, to get the frequent words, we simply used freq = nltk.FreqDist(words) to achieve our goal. For the version without stop words, we used list method to sort out punctuation and digits, and then removed stop words and meaningless words such as ("https","co","rt","http" "https","co","rt", "theyre",'de').
- Applied Stemming and Lemmatization methods to find families of derivationally related words with similar meanings. Then tokenize text file with nltk and find the top 10 most frequent words.

Find the ten most popular hashtags (#hashtag)

• To find the ten most popular hashtags, we used <u>t.startswith('#')</u> to pick up all the hashtags and then applied the same method as before when we found ten most popular words.

Find the ten most frequently mentioned usernames

• First, we set <u>cba=[t for t in word3 if t.startswith('@')]</u> to find all the mentioned users name, the used <u>freq = nltk.FreqDist(cba) to find</u> the ten most frequently mentioned usernames and printed them out.

Find the most influential tweet

- We defined that the influence score is the sum of retweet count, reply count, favorite, and quote count.
- 4 lists were created, and in each list, we applied list.append() method to find all the 4 key components retweet count, reply count, favorite, and quote count. During this process we found that if one tweet was already a retweet, then the

retweet_count would be 0, and we were supposed to check the retweeted status as well.

B. [Word Cloud]

- First, install wordcloud by typing !pip install wordcloud
- Sorted out all the meaningless words and stop words in English, asked python to read the whole text used text = all_tweets and generated a word cloud image by excuting wordcloud = Wordcloud(). generate(text). In this process we displayed words in a lower case and saved the image in both PDF and PNG files.

C. [Sentiment Analysis]

- Installed textblob
- !pip install textblob
- Analyzing the sentiment score of each tweets from the perspective of polarity and subjectivity.
- For polarity and subjectivity, created two bar charts

plt.hist(pol_list, bins=10) #, normed=1, alpha=0.75)
plt.xlabel('polarity score')
plt.ylabel('sentence count')
plt.grid(True)
plt.savefig('polarity.pdf')
plt.show()

INSIGHT:

Data-driven insights

For Microsoft(MS), the most popular and meaningful keywords(from word-cloud, keywords hashtags data) beside microsoft itself are:

- 1. Microsoft product related: Azure, Adobe, Office, xbox, xboxone, xboxonex
- 2. Function of MS Product: Cloud, service, game
- 3. General term related to MS: tech, business, buy, AI
- 4. MS People: CEO
- 5. Rivals: Amazon
- 6. Emotional words: Awful
- 7. Event related: Adobesummit

From time trend we see on the second day and third day, the "awful" word appeared on the word cloud and the polarity score dropped a little due to this event. But beyond that there is no much change over the 3-days period.

Generally, the subjectivity histogram seems behave the same, around 1300 tweets are around 0, all the other tweets are more or less subjective, which is normal compared to other tweet data. The polarity histogram is normal too, majority falls under the range 0.00-0.25, which means major tweets about MS are a little bit positive but not so much. Other finding is, beyond the majority, most tweets about MS is positive.

For Amazon, the most popular and meaningful keywords (from word-cloud, keywords hashtags data) beside Amazon itself are:

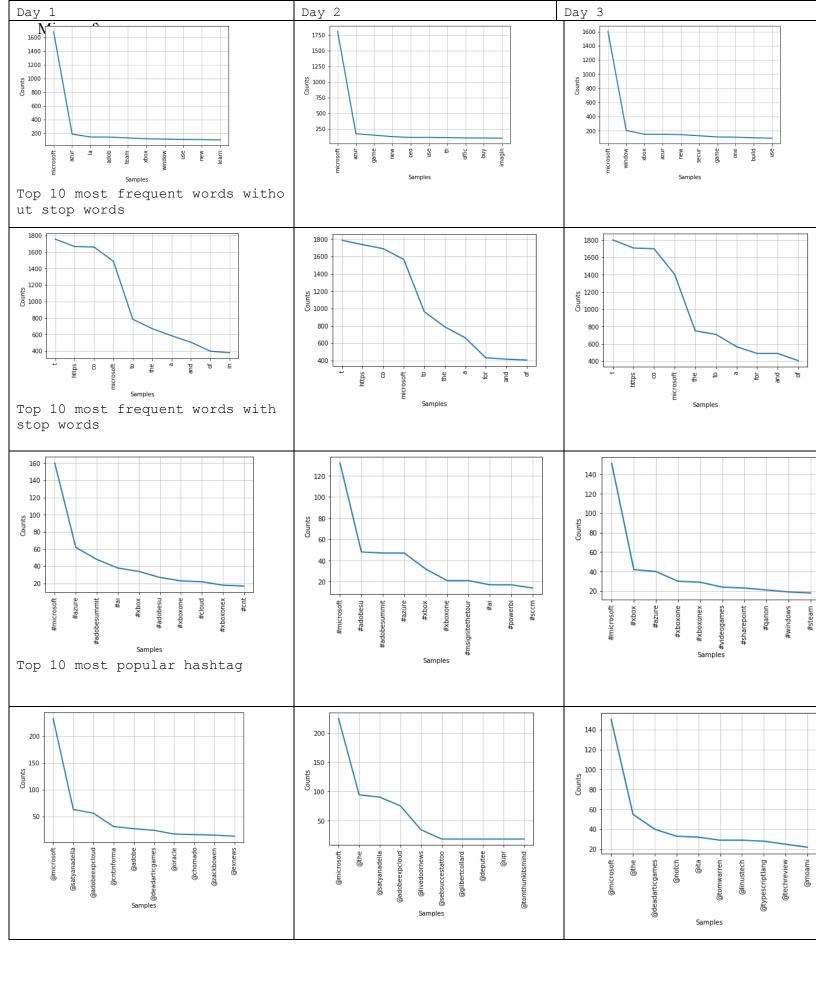
- 1. Amazon product related: drink, chips
- 2. Function of Amazon Product: Pay
- 3. General term related to Amazon: PSA (Public social announcement)
- 4. Amazon People: @divblita- the person who promote his product on amazon and twitter
- 5. Amazon partners: Yucho

Unlike microsoft tweeting info, amazon's data involved a lot of advertisement and promotion instead of brand related data. We think this makes sense as lots of small vendors online depends on amazon to sell their goods, versus microsoft does not have this distribution of merchandise like amazon. That is why there are some tweets that has a strong subjectivity compared to microsoft tweets data. Generally, the subjectivity histogram seems behave the same, around 1500 tweets are around 0, all the other tweets are more or less subjective, which is normal compared to other tweet data. The polarity histogram is normal too, majority falls under the range 0.00-0.25, which means major tweets about amazon are a little bit positive but not so much. Other finding is, beyond the majority, most tweets about amazon is positive. We think this is because there are more ads on twitter that mentioned amazon, they are more likely presenting themselves as positive. Even though we cannot conclude that amazon is well received in public eyes, we could say the market power is strong as lots of vendors rely heavily on amazon.

Project insights:

We learned a bit of the taste of managing large code project, with systematic approaches. We wrote functions as a set up and as a helper for the main part of the code to fulfill the requirements of this project. Building the helper function is so useful when it is a complex project that we can use them to break down steps and procedures. We feel building those helper functions really let us have an understanding of the purpose of functions. We feel confident about coding after this project, and we feel proud and surprised that, with resources, we can build something meaningful and insightful. Last but not least, we want to continue to explore data as a career and an interest.

Day 1	Day 2	Day 3
The most frequently tweeting	The most frequently tweeting p	The most frequently tweeting
person is LabVIEWopenJS	erson is LabVIEWopenJS	person is Rudi A.R.
the most influential tweet i	the most influential tweet i	the most influential tweet i
s this:	s this:	s this:
RT @link0230: 新型ウイルスに感	RT @i_winnn: แป็นพิมพ์ลัด สำหรับ Micr	RT @CollegeStudent: using m icrosoft word
染しました。	osoft Word	
皆さんも注意してください。	Ctrl+U = ขีดเส้นใต้ข้อความ Ctrl+F = เปิดกล่องค้นหาในบานหน้าต่	*moves an image 1 mm to the left*
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1600	1600	1600
1400	1700	1300
800 - 800 -	900 - 900 -	8 1000
E 800	800	800
% eoo -	g. 600	ğ 600 -
400	400	400
200	200	200
-1.00 -0.75 -0.50 -0.25 0.00 0.25 0.50 0.75 1.0 polarity score	0 -1.00 -0.75 -0.50 -0.25 0.00 0.25 0.50 0.75 1.0 polarity score	0 -1.00 -0.75 -0.50 -0.25 0.00 0.25 0.50 0.75 1. polarity score
The average subjectivity is: 0.2621723377606191 The average polarity is: 0.075898894 26843767	The average subjectivity is: 0.2561957499098123 The average polarity is: 0.057487888 746091995	The average subjectivity is: 0.2603240073776537 The averag e polarity is: 0.079605409400 59162



Day 1	Day 2	Day 3
The most frequently tweeting	The most frequently tweeting	The most frequently tweeting
person is 日テレ公式@宣伝部	person is 日テレ公式@宣伝部	person is 日テレ公式@宣伝部
the most influential tweet i s this: RT @bts_bighit: #BTS MAP OF THE SOUL : PERSONA Pre-orde r Notice ▶Big Hit Shop(US ONLY): htt ps://t.co/eThX79aiG7 ▶Amazon: https://t.co/Rwd	the most influential tweet i s this: RT @divblita: PSA to all my ladies! You can get these o n amazon. They're called dri nk chips. Stay safe this sum mer https://t.co/bDwrvvCm C3	the most influential tweet is this: RT @MrBeastYT: I'm going to dm someone who retweets this tweet a \$1,000 Amazon Gift card in 72 hours. If you're picked but don't follow me,
chips stay safe summer theyre called 150 d 1 V 0 1 ta 1 p S a mazon theyre stay safe summer theyre called 150 d 1 p S a mazon theyre stay safe summer theyre called drink 125 p S a 120 d 1 p S a 120	objective they be called drink they be called drink they be called drink they be called drink they be called the called they be called the called they be called the called they be called the called they be called the	o divblita psa 50 divblita psa 50 divblita psa 50 divblita psa 100 dips sley 100 dips sley
1600 1400 1200 1000 1000 1000 1000 1000 10	1400 1200 1000	1600 1400 1200 1000
1750 1500 1500 1500 1500 1500 1500 1500	The average subjectivity is: 0.21174672243266002 The avera ge polarity is: 0.14507616428 10795	The average subjectivity is: 0.18035051988135312 The average polarity is: 0.09582735171 536601

