NLP Assignment #3 - Luke Schwenke

```
In []: import pandas as pd
   import nltk
   from nltk.util import ngrams
   from nltk.tokenize import word_tokenize
   from scipy.spatial.distance import jaccard
   from nltk import TweetTokenizer
   from nltk.stem import WordNetLemmatizer

pd.set_option('display.max_rows', 100)
   pd.set_option('display.max_columns', None)
   pd.set_option('display.max_colwidth', 500)
```

Determine which news articles (news_df) are similar to each other and which tweets (tweets_df) are more similar to each other. In order to accomplish this you need to create n-grams and compare the similarity of the text using Jaccard distance.

Read news data

```
In []: news_path = 'https://storage.googleapis.com/msca-bdp-data-open/news/nlp_a_3_
    news_df = pd.read_json(news_path, orient='records', lines=True, encoding='ut
    print(f'Sample contains {news_df.shape[0]:,.0f} news articles')
    news_df.head(2)
```

Sample contains 1,018 news articles

url Out[]: date language title t Lml Bikes Car Dealers -Bikes Showrooms in Ir Explore Friday, 21 Janu 2022 Log in/Sign SearchNotifications SectionsAuto News NewsBike NewsLatestA NewsPhotosVideosElec VehiclesTrendinc Lml Bikes ReadsOffersnewF Car Dealers https://auto.hindustantimes.com/lml-2022carsFind bikesComp 0 - Lml Bikes bikes/dealers/bodh-gaya 01-21 carsCompare bikes Showrooms calculatorDealersExpl in India AutoAbout UsCont UsSITEMAPRSSTerm: UsePrivacy PolicyCopyri © HT Media Limited All ric reserved.HomeOffersnewF carsFind bikesComp carsCompare bikes calculatorDealers NewsB Pure Ev Bikes Car Deale Pure Ev Bikes Showroom India Explore Friday January 2022 Log in/Sign SearchNotifications Sections Auto News NewsBike NewsLatestA Pure Ev NewsPhotosVideosElec Bikes Car VehiclesTrending Dealers -ReadsOffersnewF https://auto.hindustantimes.com/pure-Pure Ev carsFind bikesComp en ev-bikes/dealers/avadi **Bikes** carsCompare bikes **Showrooms** calculatorDealersExpl in India AutoAbout UsCont UsSITEMAPRSSTerm: UsePrivacy PolicyCopyri © HT Media Limited All ric reserved.HomeOffersnewF carsFind bikesComp carsCompare bikes calculatorDealersCa

Read Tweets data

```
In [ ]: tweets_path = 'https://storage.googleapis.com/msca-bdp-data-open/tweets/nlp_
    tweets_df = pd.read_json(tweets_path, orient='records', lines=True)
    print(f'Sample contains {tweets_df.shape[0]:,.0f} tweets')
    tweets_df.head(2)
```

Sample contains 1,020 tweets

Out[]:		id	lang	date	name	retweeted	text
	0	1491880241782005777	en	2022- 02-10	Sir Botalot dropping the Mike!		@singervehicles Will you do a straight swap for my Renault Twingo 1.2 Extreme? \n\nThat is awesome!
	1	1501237946590142469	en	2022- 03- 08	Sabrina Ghio	RT	Australian GP Qualifying, Melbourne, 8th March 1997. #F1 The Grid ↓\n\nRow 3:\n5. Eddie Irvine (Ferrari) +2.512s\n6. Mika Hakkinen (McLaren-Mercedes) +2.602s\n\nRow 4:\n7. Johnny Herbert (Sauber-Petronas) +2.918s\n8. Jean Alesi (Benetton-Renault) +3.224s https://t.co/i4IC4LzVVN

Cleaning / Preparation

```
In [ ]: nltk.download('stopwords')
        stop words = set(stopwords.words('english'))
        import re
        def clean text(text):
            # Remove mentions
            text = re.sub(r'@[A-Za-z0-9_]+', '', text)
            # Remove hashtags (but keep the text after #)
            text = re.sub(r'#', '', text)
            # Remove RT (retweet symbol)
            text = re.sub(r'RT[\s]+', '', text)
            # Remove hyperlinks
            text = re.sub(r'https?:\/\\S+', '', text)
            # Remove newline characters
            text = re.sub(r'\n', '', text)
            # Remove carriage return characters
            text = re.sub(r'\r', '', text)
            # Remove "&"
            text = re.sub(r'&', '', text)
            # Remove other special characters and numbers
            text = re.sub(r'[^A-Za-z\s]', '', text)
            # Optionally, remove single characters (mostly left from removing mentic
            text = re.sub(r'\s+[a-zA-Z]\s+', ' ', text)
            # Convert multiple spaces to a single space
            text = re.sub(r'\s+', '', text)
            # Optionally, convert to lowercase
            text = text.lower()
            # Remove stopwords
            text = ' '.join([word for word in text.split() if word not in stop words
            return text.strip()
        [nltk data] Downloading package stopwords to
        [nltk data]
                        /Users/lmschwenke/nltk data...
        [nltk_data] Package stopwords is already up-to-date!
In []: tweets df['text clean'] = tweets df['text'].apply(clean text)
        news_df['text_clean'] = news_df['text'].apply(clean text)
        news df['title clean'] = news df['title'].apply(clean text)
In [ ]: # Examine a cleaned sample tweet
        tweets_df['text_clean'][0]
        'straight swap renault twingo extreme awesome'
Out[]:
In [ ]: # Examine a cleaned sample news article
        news_df['text_clean'][0][0:100]
        'lml bikes car dealers lml bikes showrooms india explore friday january log
Out[]:
        insign searchnotification'
```

```
In []: news_df['title_clean'][0]
Out[]: 'lml bikes car dealers lml bikes showrooms india'
In []: # Create clean dataset copies
    news_df_c = news_df
    tweets_df_c = tweets_df
```

Define a function that will create n number of ngram columns

Section #1: Tweets

Start by tokenizing the tweets

```
In []:
    def tweet_tokenize(text):
        tokenizer = TweetTokenizer()
        tokens = tokenizer.tokenize(text)
        return tokens
```

Lemmatize the tokens to the root word

```
In []: tweets_df_c['tokens'] = tweets_df_c['text_clean'].apply(tweet_tokenize)

# Lemmatize tokens
lemmatizer = WordNetLemmatizer()
tweets_df_c['tokens'] = tweets_df_c['tokens'].apply(lambda tokens: [lemmatizetweets_df_c.head(5))
```

:		id	lang	date	name	retweeted	text	
	0	1491880241782005777	en	2022- 02-10	Sir Botalot dropping the Mike!		@singervehicles Will you do a straight swap for my Renault Twingo 1.2 Extreme? \n\nThat is awesome!	straiç
	1	1501237946590142469	en	2022- 03- 08	Sabrina Ghio	RT	Australian GP Qualifying, Melbourne, 8th March 1997. #F1 The Grid ↓\n\nRow 3:\n5. Eddie Irvine (Ferrari) +2.512s\n6. Mika Hakkinen (McLaren-Mercedes) +2.602s\n\nRow 4:\n7. Johnny Herbert (Sauber- Petronas) +2.918s\n8. Jean Alesi (Benetton-Renault) +3.224s https://t.co/i4IC4LzVVN	qualit th ed m rov saub
	2	1505982695129718784	en	2022- 03-21	Colin N. Walker Walker FBPE	RT	#BoycottRenault\n\nThink of the blood of thousands of Ukrainian women and children pouring from every Renault car. \n\nhttps://t.co/rbU01Sy9DU	boyc k u c
	3	1516744110463463426	en	2022- 04- 20	Yvette Lissman	RT	Almost 200,000 workers in Russia still on western payrolls\n\nMcDonald's, IKEA, Renault, Levi Strauss, & Description and their spay salaries to thousands of their employees while their operations in are suspended\nCoca-Cola, Yum Brands, KFC didn't confirm if they still pay shttps://t.co/KWDhCo1dM0	rus pay str sal susp yum c
	4	1493777143347630086	en	2022- 02-16	Andile Xaba ⊵	RT	Take a selfie with the New #Renault #ClioV and WIN R1000 fuel voucher @tableviewrenault & tag @tableviewrenault & tamp; #renaulttableviewcliov random winner announced 28.02.2022 @BradAtRenault 0825662336 to book a test drive https://t.co/D7V2GnJa8B	renal renal

Out[]

Create n number of ngrams columns based on the lemmatized token tweets

```
In []:
         for i in range(1, 11):
             tweets df c['ngrams '+str(i)] = tweets df c['tokens'].apply(create ngram
         tweets_df_c.head(1)
In [ ]:
Out[ ]:
                             id lang
                                        date
                                                name retweeted
                                                                          text text_clean
                                                                                             to
                                                                 @singervehicles
                                                                   Will you do a
                                                                                  straight
                                                                                           str
                                                  Sir
                                                                   straight swap
                                                                                    swap
                                               Botalot
                                      2022-
                                                                  for my Renault
                                                                                   renault
                                                                                            re
         0 1491880241782005777
                                   en
                                             dropping
                                       02-10
                                                                     Twingo 1.2
                                                                                   twingo
                                                                                            t٧
                                                  the
                                                                      Extreme?
                                                                                  extreme
                                                                                            ext
                                                Mike!
                                                                     \n\nThat is
                                                                                 awesome awes
                                                                   awesome! 🤚
In []: from nltk.metrics import jaccard distance
         def calculate jaccard distance(ngrams1, ngrams2):
             # Convert n-gram lists to sets
             set_ngrams1 = set(ngrams1)
             set_ngrams2 = set(ngrams2)
```

distance = 1 - jaccard distance(set ngrams1, set ngrams2)

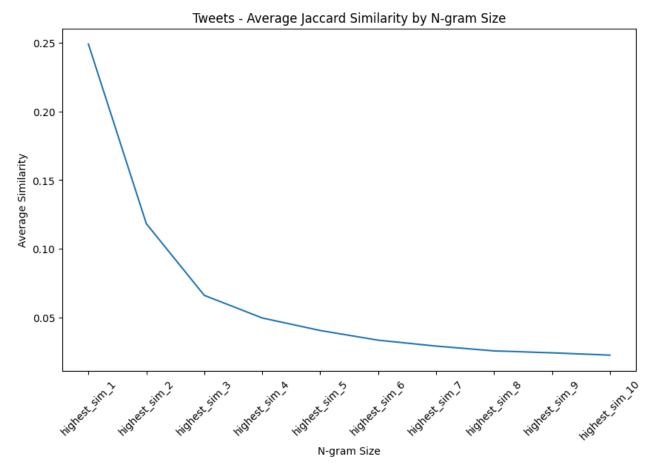
return distance

```
In [ ]: def compute max jaccard(df, num):
            # Create new columns for storing the highest Jaccard similarity and corr
            df['highest sim '+str(num)] = 0.0 # Initialize with 0.0 (no similarity)
            df['highest sim ind '+str(num)] = -1 # Initialize with -1 (no index)
            from itertools import combinations
            # Iterate over all unique pairs of rows in the dataframe
            for i, j in combinations(df.index, 2):
                # Calculate Jaccard similarity, handle cases with empty n-grams
                if len(df['ngrams_'+str(num)][i]) == 0 or len(df['ngrams_'+str(num)]
                    similarity = 0 # If either n-gram set is empty, similarity is 0
                else:
                    # Calculate Jaccard similarity for non-empty n-gram sets
                    similarity = calculate jaccard_distance(df.at[i, 'ngrams_'+str(n
                # Ignore perfect matches (similarity = 1), considering them as zero
                if similarity == 1:
                    similarity = 0
                # Update the record for the highest similarity for each row
                # If current similarity is higher than the stored value, update it a
                if similarity > df.at[i, 'highest sim '+str(num)]:
                    df.at[i, 'highest sim ind '+str(num)] = j # Update index
                    df.at[i, 'highest sim '+str(num)] = similarity # Update similar
                if similarity > df.at[j, 'highest_sim_'+str(num)]:
                    df.at[j, 'highest sim ind '+str(num)] = i # Update index for th
                    df.at[j, 'highest sim_'+str(num)] = similarity # Update similar
            # At the end of this function, each row in df will have the highest Jaco
            # and the index of the row with which this highest similarity is achieve
In []: for i in range(1,11):
            compute max jaccard(tweets df c, i)
In [ ]: sim columns
        Index(['highest sim 1', 'highest sim 2', 'highest sim 3', 'highest sim 4',
Out[ ]:
               'highest_sim_5', 'highest_sim_6', 'highest_sim_7', 'highest_sim_8',
               'highest sim 9', 'highest sim 10'],
              dtype='object')
```

```
In []: import matplotlib.pyplot as plt

sim_columns = tweets_df_c.filter(regex='^highest_sim_\d+').columns
avg_sim = tweets_df_c[sim_columns].mean()

# Plot using seaborn's lineplot
plt.figure(figsize=(10, 6)) # Adjust the figure size as necessary
sns.lineplot(data=avg_sim)
plt.xticks(rotation=45) # Rotate x-axis labels for better visibility
plt.xlabel('N-gram Size') # Set title for the x-axis
plt.ylabel('Average Similarity') # Set title for the y-axis
plt.title('Tweets - Average Jaccard Similarity by N-gram Size') # Set the toplt.show()
```



Based on the above plot, I would conclude an ngram value of 3 is best for tweets. This makes sense because ~5 is usually ideal for shorter texts whereas a higher number like 10 is better for long articles / books. So tweets aligning with n=3 seems correct.

```
In [ ]: tweets_df[['highest_sim_3','highest_sim_ind_3','text_clean','text']].nlarges
```

highest_sim_3 highest_sim_ind_3 text_clean Out[]: text renault kiger stunning yet muscular suv stance Renault Kiger is stunning yet crafted complement muscular SUV stance is free spirit renault crafted to complement your cars nagercoil free spirit. \n#renault #cars 15 0.947368 378 sportysmart #nagercoil #Sportysmart renaultcars #renaultcars #renaultindia renaultindia #bestcars #buycarsnagercoil bestcars #morespacing #renaultkiger buycarsnagercoil https://t.co/0gaRyxca7h morespacing renaultkiger kiger stunning yet muscular suv stance https://t.co/DFfqSVOypZ crafted complement \nKiger is stunning yet free spirit renault muscular SUV stance is cars nagercoil crafted to complement your sportysmart free spirit.\n#renault #cars 378 0.947368 15 renaultcars #nagercoil #Sportysmart renaultindia #renaultcars #renaultindia bestcars #bestcars #buycarsnagercoil buycarsnagercoil #morespacing #renaultkiger morespacing https://t.co/LzLJ0ljpyT renaultkiger CNN: After Renault cnn renault announced its departure this announced departure week week from the Russian auto russian auto market market amid the country's amid countrys war war with Ukraine, Moscow's 181 0.909091 877 mayor announced its factory ukraine moscows will be used to restart the mayor announced factory used restart defunct Soviet-era Moskvich defunct sovietera car brand. moskvich car brand https://t.co/QtlpxQ9d6l (via https://t.co/lgfjwGPwdG) via renault announced After Renault announced its departure week departure this week from the russian auto market Russian auto market amid the amid countrys war country's war with Ukraine, 877 0.909091 181 ukraine moscows Moscow's mayor announced mayor announced its factory will be used to factory used restart restart the defunct Soviet-era defunct sovietera Moskvich car brand. moskvich car brand https://t.co/F3pRGrtYQX reuters french Reuters: French automaker renault resumes production automaker Renault resumes moscow march production in Moscow.\n\nOn company decided March 22, the company resume operations decided to resume operations 140 0.846154 659 country renault in the country. Renault is the

			majority owner avtovaz russias largest car manufacturer famous lada brand	majority owner of AvtoVaz, Russia's largest car manufacturer famous for the Lada brand.
241	0.846154	659	reuters french automaker renault resumes production moscow march company decided resume operations country renault majority owner avtovaz russias largest car manufacturer famous lada brand	Reuters: French automaker Renault resumes production in Moscow.\n\nOn March 22, the company decided to resume operations in the country. Renault is the majority owner of AvtoVaz, Russia's largest car manufacturer famous for the Lada brand.
242	0.846154	659	reuters french automaker renault resumes production moscow march company decided resume operations country renault majority owner avtovaz russias largest car manufacturer famous lada brand	Reuters: French automaker Renault resumes production in Moscow.\n\nOn March 22, the company decided to resume operations in the country. Renault is the majority owner of AvtoVaz, Russia's largest car manufacturer famous for the Lada brand.
406	0.846154	659	reuters french automaker renault resumes production moscow march company decided resume operations country renault majority owner avtovaz russias largest car manufacturer famous lada brand	Reuters: French automaker Renault resumes production in Moscow.\n\nOn March 22, the company decided to resume operations in the country. Renault is the majority owner of AvtoVaz, Russia's largest car manufacturer famous for the Lada brand.
430	0.846154	659	reuters french automaker renault resumes production moscow march company decided resume operations country renault majority owner avtovaz russias largest car manufacturer famous lada brand	Reuters: French automaker Renault resumes production in Moscow.\n\nOn March 22, the company decided to resume operations in the country. Renault is the majority owner of AvtoVaz, Russia's largest car manufacturer famous for the Lada brand.

The highest similarity tweets are related to the french automatker, Renault cars, and mentions of Moscow.

In []:	<pre>non_zero_tweets = tweets_df[tweets_df['highest_sim_3'] > 0] non_zero_tweets[['highest_sim_3','highest_sim_ind_3', 'text_clean', 'text']]</pre>					
Out[]:		highest_sim_3	highest_sim_ind_3	text_clean		
	667	0.019231	455	jean alesi finished nd first podium points since moving benettonrenault michael schumacher also took first podium finish points ferrari driver rd lap behind race winner damon hill brazilian gp interlagos st march	Jean Alesi finished 2nd, his first & points with since more Benetton-Renault. Michael Schumach took his first podium finish & p a Ferrari driver with 3rd in his F31 behind race winner Damon Hill. Brazi Interlagos, 31st March 19 https://t.co/leJw	
	664	0.022222	898	rothmans williamsrenault launch estoril chassis fw engine renault rs v tyres goodyear eagle damon hill jacques villeneuve test driver jeanchristophe boullion	Rothmans Williams-Renault Launch, 1996. #F1 ■ InChassis: FW18\n Renault RS8, 3.0L V10\nTyres: Go Eagle F1\n5. Damon Hill ■\n6. S Villeneuve InTest Driver: Jean-Chr Boullion Inttps://t.co/gi06	
				mild seven renault car launch event monaco chassis engine mecachromebuilt	Mild Seven Renault F1 2005 car launc at Monaco\n\nChassis R25\ı Mecachrome-built Renault RS25 3.	

898	0.022222	664	renault rs v naturally aspirated power hp rpm weight kg tyres michelin points race wins podiums world titles elplan	72° naturally aspirated\nPower 800- @ 19,000 RPM\nWeight 605 kc Michelin\n\n191 points, 8 race of podiums and 2 World Titles\n\n https://t.co/L770l
641	0.023810	877	car news today mr va auto legendary sovietera moskvich car could revived renault exit mr va auto car buying hero autoblog news car check autoblogs news	Car News of Today Auto\n.\n'Legendary' Soviet-era M car could be revived after Renault exi VA Auto - Your Car Hero\nhttps://t.co/MWNhcYg5LC\n#A#Car\nhttps://t.co/tNfYThlk8q\nchec AutoBlogs news here\nhttps://t.co/fpt https://t.co/unQ3
771	0.023810	1012	michael schumacher mild seven benettonrenault lap qualifying couldnt beat session best german claimed pole position spanish gp qualifying barcelona th may	Michael Schumacher (Mild Seven Be Renault B195) on a lap in qualifyir couldn't beat his session best German claimed Pole Position.\nSpa Qualifying, Barcelona, 13th Ma #F1\nhttps://t.co/DMcxB
849	0.023810	61	tailored meet demands uk professionals requiring highspec vehicle reflects business values lifestyle new renault trucks trafic red edition tonnes prioritises productivity safety driver comfort renaulttrucks	Tailored to meet the demand professionals requiring a high-spec that reflects their own business validifestyle, the new Renault Trucks Transcription from 2.8 - 3.1 tonnes productivity, safety and driver com \n\n#RenaultTrucks https://t.co/jM69
21	0.026316	718	behind powerfully elegant suv silhouette hides technological marvel allnew renault austral almost ready closeup	Behind this powerfully elegal silhouette hides a technological mar new Renault #Austral is almost read close-up! #comingsoon\nLear https://t.co/eMQE(https://t.co/ETN)

comingsoon learn

60	0.026316	200	riccardo patrese brought williamsrenaultfwc home rd position italians fourth consecutive podium finish french grand prix paul ricard july motorsport images	Riccardo Patrese brought his W Renault-FW12C home in 3rd positior was the Italian's fourth consecutive finish. \n\nFrench Grand Prix, Paul R July 1989.\n\n© Motorsport Images https://t.co/w00
61	0.026316	822	today cest participate tunemyt challenge share beautiful renault trucks high pimped truck within new renault trucks dlc win incredible suprises merch eshop	You have until today 12am (participate in the #Ti challenge.\nShare you most b Renault Trucks T & Dight pimpe within the new Renault Trucks @SCSsoftware\n to win some incomparises from our merch https://t.co/I03MpE
718	0.026316	21	allnew renault austral sound quality renault group brand news renault follow mycarnewsonline see latest news reviews tech french manufacturer right car news watch full video youtube fol	The all-new Renault Austral: The s quality Renault Group\n\nFor mor news from Renault follow MyCarNews See the latest news, reviews and te the French manufacturer right here Car News.\n\nWatch Full V YouTube\n\nFol\nhttps://t.co/vqXN

The lowest similarity rows, as expected do not appear related. For example, the first few tweets talk about Renault and Ricciardo and do not share as much simmilar text/language

Section #2: Articles

```
In []: def news_tokenize(text):
    # Directly tokenize the text using the word_tokenize function
    tokens = word_tokenize(text)
    return tokens
```

```
In []: # Apply the cleaning functio nto the column that we defined earlier
    news_df_c['text_clean'] = news_df_c['text'].apply(clean_text)
    news_df_c['title_clean'] = news_df_c['title'].apply(clean_text)

In []: # Tokenize the news articles text and titles
    news_df_c['text_tokens'] = news_df_c['text_clean'].apply(news_tokenize)
    news_df_c['title_tokens'] = news_df_c['title_clean'].apply(news_tokenize)

In []: # Lemmatize the tokens to the root
    lemmatizer = WordNetLemmatizer()
    news_df_c['text_tokens'] = news_df_c['text_tokens'].apply(lambda tokens: [lenews_df_c['title_tokens'] = news_df_c['title_tokens'].apply(lambda tokens: [news_df_c['text_tokens'] = news_df_c['title_tokens'].apply(lambda tokens: [news_df_c['text_tokens', 'title_tokens']].head(5)
```

Out[]: title_tokens text_tokens [lml, bike, car, dealer, lml, bike, showroom, india, explore, friday, january, log, insign, searchnotificationstop, sectionsauto, newscar, newsbike, [lml, bike, car, newslatestauto, newsphotosvideoselectric, vehiclestrendingmy, dealer, Iml, readsoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, 0 bike, calculatordealersexplore, autoabout, uscontact, ussitemaprssterms, showroom, useprivacy, policycopyright, ht, medium, limited, right, india] reservedhomeoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, calculatordealerscar... [pure, ev, bike, car, dealer, pure, ev, bike, showroom, india, explore, friday, january, log, insign, searchnotificationstop, sectionsauto, newscar, newsbike, [pure, ev, bike, newslatestauto, newsphotosvideoselectric, vehiclestrendingmy, car, dealer, readsoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, 1 pure, ev, bike, calculatordealersexplore, autoabout, uscontact, ussitemaprssterms, showroom, useprivacy, policycopyright, ht, medium, limited, right, india] reservedhomeoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, calculator... [syncron, price, [syncron, price, selected, mitsubishi, motor, corporation, boost, enhanced, selected, service, part, pricing, strategyskip, contentcircle, country, music, mitsubishi, lifestyleadvertise, usteacher, tributeask, expertthank, nursebe, excellentwatch, motor, livenewselectionsvaccine, 2 corporation, trackervideoweathersportscommunitycontestsabout, uscovidsearchhomesee, boost. snap, send, itnewsstorm, centurynationalstateeditorialinvestigateeast, texas, enhanced, ag, newscrimeeast, texas, nowthe, next, normalsept, thweathersign, service, part, thundercalllake, levelsproje... pricing, strategy] [mahindra, tease, future, electric, lineup, motoroids, motoroids, blogmotoroids, forum, authorscontact, ussubmit, storyadvertise, usprivacy, policy, search, [mahindra, homeauto, newsfeatureslaunchesupcoming, carsupcoming, suvsupcoming, tease, future, 3 bikesrecent, launchesreviewsmodsmodified, bikesmodified, carsinteresting, electric, lineup, offbeatlistscc, bikescc, bikes motoroids] cc, bikesbikes, indiacars, indiaforums, trending, triumph, trident, get, pricier, homenewsmahindra, tease, future, ele... [jawa, bike, car, dealer, jawa, bike, showroom, india, explore, saturday, january, log, insign, searchnotificationstop, sectionsauto, newscar, newsbike, [jawa, bike, newslatestauto, newsphotosvideoselectric, vehiclestrendingmy, car, dealer, readsoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, 4 jawa, bike, calculatordealersexplore, autoabout, uscontact, ussitemaprssterms, showroom, useprivacy, policycopyright, ht, medium, limited, right, india] reservedhomeoffersnewfind, carsfind, bikescompare, carscompare, bikesemi, calculatordealer...

Examine the most common tokens

```
In [ ]: from collections import Counter
Counter(news_df_c['text_tokens'].explode()).most_common(10)
```

```
Out[]: [('petrol', 15661),
          ('lakh', 13718),
          ('mt', 9260),
          ('offer', 8434),
          ('turbo', 8338),
          ('car', 6568),
          ('complete', 5837),
          ('cc', 5216),
          ('xv', 5113),
          ('aprview', 5053)]
        Counter(news_df_c['title_tokens'].explode()).most_common(10)
Out[]: [('car', 555),
          ('bike', 336),
          ('offer', 255),
          ('discount', 254),
          ('march', 254),
          ('dealer', 251),
          ('showroom', 247),
          ('india', 223),
          ('news', 95),
          ('tata', 78)]
```

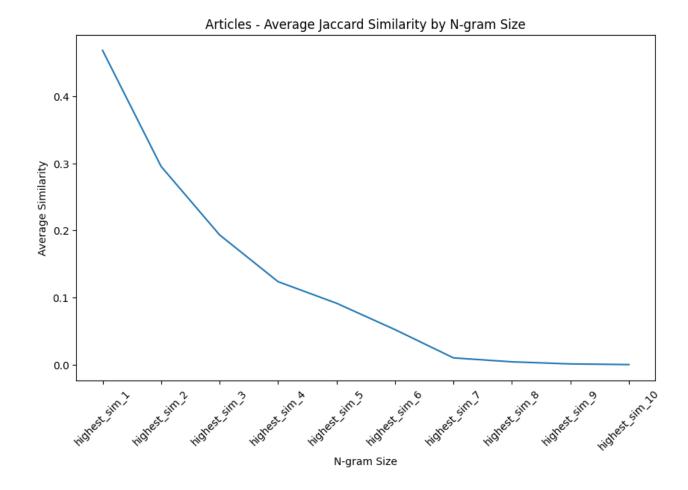
The title tokens appear more valuable with less noise -- we will use these title tokens for the remaining analysis

Out[]: url date language title te:

https://auto.hindustantimes.com/lml- 2022bikes/dealers/bodh-gaya 01-21 Lml Bikes Car Dealers en - Lml Bikes Showrooms in India

Lml Bikes Car Dealers - Lr Bikes Showrooms in Ind Explore Friday, 21 Janua 2022 Log in/Sign ι SearchNotificationsTo SectionsAuto NewsC NewsBike NewsLatestAu NewsPhotosVideosElectr VehiclesTrendingN ReadsOffersnewFir carsFind bikesCompa carsCompare bikesEN calculatorDealersExplo AutoAbout UsConta UsSITEMAPRSSTerms UsePrivacy PolicyCopyrig © HT Media Limited All righ reserved.HomeOffersnewFir carsFind bikesCompa carsCompare bikesEN calculatorDealersC NewsBik

```
for i in range(1,11):
In [ ]:
            compute max jaccard(news df c, i)
In [ ]:
        import matplotlib.pyplot as plt
        sim columns = news df c.filter(regex='^highest sim \d+').columns
        avg sim = news df c[sim columns].mean()
        # Plot using seaborn's lineplot
        plt.figure(figsize=(10, 6)) # Adjust the figure size as necessary
        sns.lineplot(data=avg sim)
        plt.xticks(rotation=45) # Rotate x-axis labels for better visibility
        plt.xlabel('N-gram Size') # Set title for the x-axis
        plt.ylabel('Average Similarity') # Set title for the y-axis
        plt.title('Articles - Average Jaccard Similarity by N-gram Size')
                                                                            # Set the
        plt.show()
```



Based on the above plot, I would conclude an ngram value of 4 is best for the article titles. I will continue with this n value going forward.

In []:	news	_df_c[['highe	st_sim_4','highe	st_sim_ind_4	','title_clean', 'text_clean']]
Out[]:		highest_sim_4	highest_sim_ind_4	title_clean	
	6	0.857143	349	exnissan us exec kelly gets suspended sentence go home	exnissan us exec kelly gets suspende news us us politics politics world world co
	349	0.857143	6	exnissan us exec kelly gets suspended sentence go home ourquadcities	exnissan us exec kelly gets suspende termprimary menunewslocal newslocal
				france issues arrest	

35	0.833333	760	warrant disgraced auto tycoon ghosn	newsbusinesseducationweathertra
534	0.833333	760	france issues arrest warrant disgraced auto tycoon ghosn	france issues arrest warrant di uscirculationnewslett
664	0.833333	790	russias war spurs corporate exodus exposes business risks	russias war spurs corporate exodu usadvertisesubscribeprivacy noticeterms
760	0.833333	35	france issues arrest warrant disgraced auto tycoon ghosn kesq	buttonchevronrightchevronleftchevronups valley questions answerededucationiteam
790	0.833333	664	russias war spurs corporate exodus exposes business risks khon	russias war spurs corporate exodus expos menunewshawaii new
800	0.800000	912	japan prosecutors appeal exnissan executive kelly trial	japan prosecutors appeal ex regionweatherschool closingsobitua
912	0.800000	800	japan prosecutors appeal exnissan executive kelly trial wtmj	japan pr homenewslocalnationalcoronavirust
294	0.714286	35	france issues arrest warrant disgraced auto tycoon ghosn knwa fox	france issues arrest warrant di menunewsknwafoxaround arkansasruss

The highest similarity news articles by their title primarily are centered around france and nissans. A lof of the language between them is expectedly similar.

In []:	<pre>non_zero_articles = news_df_c[news_df_c['highest_sim_4'] > 0]</pre>	
	non_zero_articles[['highest_sim_4','highest_sim_ind_4', 'title_clean', 'te	xt

Out[]:		highest_sim_4	highest_sim_ind_4	title_clean	text_clean
	186	0.041667	396	pm gramin digital saksharta abhiyaan renault india partners csc egovernance services support pm gramin digital saksharta abhiyaan auto news et auto	pm gramin digital saksharta abhiyaan renault india partners csc egovernance services support pm gramin digital saksharta abhiyaan auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose igno
	237	0.047619	396	hero motocorp price hike hero motocorp hike motorcycle scooter prices inr july auto news et auto	hero motocorp price hike hero motocorp hike motorcycle scooter prices inr july auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies et aut
	185	0.050000	396	renault exports renaults made india products cross one lakh export milestone	renault exports renaults made india products cross one lakh export milestone auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see

			auto news et auto	privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies et autoa
386	0.050000	396	biofuel us biofuel industry defends record biden administration mulls policy reform auto news et auto	biofuel us biofuel industry defends record biden administration mulls policy reform auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies
492	0.050000	396	global energy transition global energy transition cause shortterm economic pain report auto news et auto	global energy transition global energy transition cause shortterm economic pain report auto news et auto etautonewsnews passenger vehicle commercial vehicle two wheelers automotive components industry tyres aftermarket policy auto technology people movement oil lubes new launches raw material financial results auto finance featuresfeatures trends autopreneur etauto tv industryspeakindustryspeak interviews autologue etauto insights dealersdata analytics detautolytics reports etau
660	0.050000	396	india fuel demand india expects fuel demand grow next fiscal year auto news et auto	india fuel demand india expects fuel demand grow next fiscal year auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies et autoanalyticsnec

lg energy solution gm sets bln

	961	0.052632	396	lg energy solution gm sets bln michigan electric vehicle plants auto news et auto	michigan electric vehicle plants auto news et auto etautonewsnews passenger vehicle commercial vehicle two wheelers automotive components industry tyres aftermarket policy auto technology people movement oil lubes new launches raw material financial results auto finance featuresfeatures trends autopreneur etauto tv industryspeakindustryspeak interviews autologue etauto insights dealersdata analyticsdata analytics etautolytics reports etautotvbrand solutionsbran
	266	0.055556	396	msme promoters aima introduces month management course msme promoters auto news et auto	msme promoters aima introduces month management course msme promoters auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies et autoanalytic
	396	0.058824	452	ev policy state governments push ahead ev road auto news et auto	ev policy state governments push ahead ev road auto news et auto updated terms conditions privacy policy click continue accept continue et autoaccept updated privacy cookie policydear user et auto privacy cookie policy updated align new data regulations european union please review accept changes continue using websiteyou see privacy policy cookie policy use cookies ensure best experience websiteif choose ignore message well assume happy receive cookies et autoanalyticsnecessarynewsletter na
				rupee value	rupee value rupee rises paise close us dollar auto news et auto etautonewsnews passenger vehicle commercial vehicle two wheelers automotive components industry tyres aftermarket policy auto

452	0.058824	396	rupee rises paise close us dollar auto	technology people movement oil lubes new launches raw material financial results auto finance
			news et auto	featuresfeatures trends autopreneur
				etauto tv
				industryspeakindustryspeak
				interviews autologue etauto insights
				dealersdata analyticsdata analytics
				etautolytics reports etautotybrand solutionsbrand solutions etauto

The most dissimilar news articles/titles do not show much relation and span various topics like india, scooter prices, and fuel

Summary: The n for tweets and articles were different. n for tweets was 3 and n for article titles was 4 but it could have been 3 as well. Even though I set them to different n values, they both were close and were smaller than 5 ngrams. This makes sense because tweets and titles in general do not have long text, so setting a 10+ ngram value could encompass the whole text itself in some instances.