sarcasm-detection-2

March 19, 2024

```
[]: mkdir -p ~/.kaggle
     !cp kaggle.json ~/.kaggle/
[]: | !kaggle datasets download -d rmisra/news-headlines-dataset-for-sarcasm-detection
    Warning: Your Kaggle API key is readable by other users on this system! To fix
    this, you can run 'chmod 600 /root/.kaggle/kaggle.json'
    Downloading news-headlines-dataset-for-sarcasm-detection.zip to /content
      0% 0.00/3.30M [00:00<?, ?B/s]
    100% 3.30M/3.30M [00:00<00:00, 122MB/s]
[]: import zipfile
     zip_ref = zipfile.ZipFile('/content/
      ⇔news-headlines-dataset-for-sarcasm-detection.zip', 'r')
     zip_ref.extractall('/content')
     zip_ref.close()
[]: import numpy as np
     import pandas as pd
     import seaborn as sns
     import matplotlib.pyplot as plt
[]: SD_DATA = pd.read_json("/content/Sarcasm Headlines_Dataset_v2.json",lines=True)
[]: SD DATA.head(20)
[]:
                                                                 headline \
         is_sarcastic
                    1 thirtysomething scientists unveil doomsday clo...
     0
     1
                       dem rep. totally nails why congress is falling...
                       eat your veggies: 9 deliciously different recipes
     2
     3
                       inclement weather prevents liar from getting t...
                       mother comes pretty close to using word 'strea...
     4
                    1
     5
                    0
                                                    my white inheritance
     6
                    0
                              5 ways to file your taxes with less stress
     7
                    1 richard branson's global-warming donation near...
     8
                       shadow government getting too large to meet in...
                    0
                                      lots of parents know this scenario
```

```
10
                  this lesbian is considered a father in indiana...
11
                  amanda peet told her daughter sex is 'a specia...
12
                  what to know regarding current treatments for ...
13
                  chris christie suggests hillary clinton was to...
14
                  ford develops new suv that runs purely on gaso...
15
                  uber ceo travis kalanick stepping down from tr...
16
                  area boy enters jumping-and-touching-tops-of-d...
               1
17
               1
                       area man does most of his traveling by gurney
18
               0
                            leave no person with disabilities behind
19
               0
                  lin-manuel miranda would like to remind you to...
```

article_link

\

```
0
    https://www.theonion.com/thirtysomething-scien...
1
    https://www.huffingtonpost.com/entry/donna-edw...
2
    https://www.huffingtonpost.com/entry/eat-your-...
3
    https://local.theonion.com/inclement-weather-p...
4
    https://www.theonion.com/mother-comes-pretty-c...
5
    https://www.huffingtonpost.com/entry/my-white-...
6
    https://www.huffingtonpost.com/entry/5-ways-to...
7
    https://www.theonion.com/richard-bransons-glob...
8
    https://politics.theonion.com/shadow-governmen...
9
    https://www.huffingtonpost.comhttp://pubx.co/6...
10
   https://www.huffingtonpost.com/entry/this-lesb...
   https://www.huffingtonpost.com/entry/amanda-pe...
11
12
   https://www.huffingtonpost.com/entry/what-to-k...
   https://www.huffingtonpost.com/entry/chris-chr...
   https://www.theonion.com/ford-develops-new-suv...
   https://www.huffingtonpost.com/entry/uber-ceo-...
15
16
   https://www.theonion.com/area-boy-enters-jumpi...
   https://local.theonion.com/area-man-does-most-...
17
   https://www.huffingtonpost.com/entry/leave-no-...
18
   https://www.huffingtonpost.com/entry/lin-manue...
```

[]: SD_DATA.tail(20)

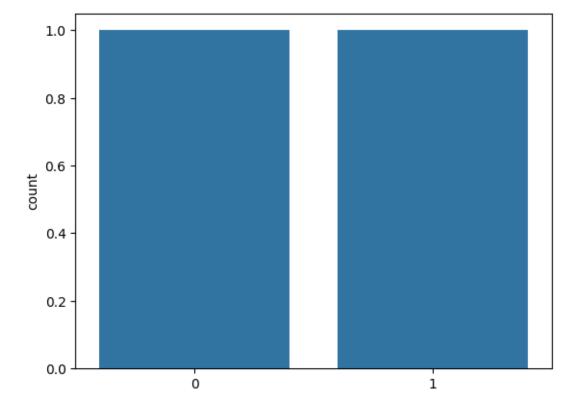
[]:	is_sarcastic	headline
28599	0	tight wisconsin house primary too close to cal
28600	0	diagnosing and curing our sick health system
28601	0	cooking off the cuff: bluefish in saor - a new
28602	0	omarosa turns on trump: wouldn't vote for him
28603	1	area eyesore also a data technician
28604	0	new york attorney general conducting 'inquiry'
28605	0	get ready to capture pokémon in the real world
28606	0	amy schumer pens letter to tampa trump fans wh
28607	0	what our grieving family needs from loved ones
28608	0	stephen colbert attempts to list everything tr
28609	0	bakery owner vows to stop making wedding cakes

```
28611
                       1
                                     grandmother doesn't care for new priest
     28612
                       1
                          polish rapper under fire for use of the word '...
     28613
                       0
                          how san antonio's dominant defense is fueling ...
     28614
                       1
                                jews to celebrate rosh hashasha or something
     28615
                       1
                          internal affairs investigator disappointed con...
     28616
                          the most beautiful acceptance speech this week...
                       0
     28617
                       1
                          mars probe destroyed by orbiting spielberg-gat...
     28618
                                          dad clarifies this not a food stop
                        1
                                                  article link
     28599
            https://www.huffingtonpost.com/entry/wisconsin...
     28600
            https://www.huffingtonpost.com/entry/diagnosin...
     28601
            https://www.huffingtonpost.com/entry/cooking-o...
            https://www.huffingtonpost.com/entry/omarosa-t...
     28602
     28603
            https://www.theonion.com/area-eyesore-also-a-d...
            https://www.huffingtonpost.com/entry/investiga...
     28604
     28605
            https://www.huffingtonpost.com/entry/pokemon-g...
     28606
            https://www.huffingtonpost.com/entry/amy-schum...
     28607
            https://www.huffingtonpost.com/entry/what-our-...
     28608
            https://www.huffingtonpost.com/entry/stephen-c...
     28609
            https://www.huffingtonpost.com/entry/jack-phil...
     28610
            https://sports.theonion.com/congressman-picked...
     28611
            https://local.theonion.com/grandmother-doesn-t...
     28612
            https://entertainment.theonion.com/polish-rapp...
     28613
            https://www.huffingtonpost.com/entry/san-anton...
            https://www.theonion.com/jews-to-celebrate-ros...
     28614
            https://local.theonion.com/internal-affairs-in...
     28615
     28616
            https://www.huffingtonpost.com/entry/andrew-ah...
            https://www.theonion.com/mars-probe-destroyed-...
     28617
     28618
            https://www.theonion.com/dad-clarifies-this-no...
     SD_DATA.shape
[]: (28619, 3)
     SD_DATA.columns
[]: Index(['is_sarcastic', 'headline', 'article_link'], dtype='object')
[]: # check the data types in the columns
     SD_DATA.dtypes
[]: is_sarcastic
                      int64
     headline
                     object
     article_link
                     object
     dtype: object
```

congressman picked last for committee on youth...

28610

```
[]: # check the null values in data
     SD_DATA.isna().sum()
[]: is_sarcastic
                     0
                     0
    headline
     article_link
                     0
     dtype: int64
[]: #checking the unique values in 'is_sarcastic' column
     SD_DATA.is_sarcastic.unique()
[]: array([1, 0])
[]: #checking the value counts in 'is_sarcastic' column
     SD_DATA.is_sarcastic.value_counts()
     sns.countplot(SD_DATA['is_sarcastic'].value_counts())
     plt.show()
```



```
[]: #drop 'article_link' column
SD_DATA = SD_DATA.drop('article_link', axis=1)
```

```
[]: #check the data
     SD_DATA.head(10)
[]:
        is_sarcastic
                                                                headline
     0
                   1 thirtysomething scientists unveil doomsday clo...
                   0 dem rep. totally nails why congress is falling...
     1
     2
                   O eat your veggies: 9 deliciously different recipes
     3
                   1 inclement weather prevents liar from getting t...
                   1 mother comes pretty close to using word 'strea...
     4
     5
                                                    my white inheritance
     6
                   0
                             5 ways to file your taxes with less stress
     7
                   1 richard branson's global-warming donation near...
     8
                   1 shadow government getting too large to meet in...
     9
                   0
                                      lots of parents know this scenario
[]: import re
     import nltk
     from nltk.corpus import stopwords
     from nltk.tokenize import word_tokenize
     from nltk.stem import WordNetLemmatizer
[]: nltk.download('punkt')
     nltk.download('stopwords')
     nltk.download('wordnet')
    [nltk_data] Downloading package punkt to /root/nltk_data...
                  Unzipping tokenizers/punkt.zip.
    [nltk_data]
    [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk_data]
                  Unzipping corpora/stopwords.zip.
    [nltk_data] Downloading package wordnet to /root/nltk_data...
[]: True
[]: # Convert to Lowercase using lower()
     SD_DATA['headline'] = SD_DATA['headline'].apply(lambda x: ' '.join(x.lower()_

→for x in x.split()))
     SD_DATA
[]:
                                                                    headline
            is_sarcastic
     0
                          thirtysomething scientists unveil doomsday clo...
     1
                          dem rep. totally nails why congress is falling...
     2
                          eat your veggies: 9 deliciously different recipes
                          inclement weather prevents liar from getting t...
     3
     4
                          mother comes pretty close to using word 'strea...
     28614
                       1
                               jews to celebrate rosh hashasha or something
     28615
                       1 internal affairs investigator disappointed con...
```

```
28616 0 the most beautiful acceptance speech this week...
28617 1 mars probe destroyed by orbiting spielberg-gat...
28618 1 dad clarifies this not a food stop
```

[28619 rows x 2 columns]

```
[]: #types of stopwords
from nltk.corpus import stopwords
", ".join(stopwords.words('english'))
```

[]: "i, me, my, myself, we, our, ours, ourselves, you, you're, you've, you'll, you'd, your, yours, yourself, yourselves, he, him, his, himself, she, she's, her, hers, herself, it, it's, its, itself, they, them, their, theirs, themselves, what, which, who, whom, this, that, that'll, these, those, am, is, are, was, were, be, been, being, have, has, had, having, do, does, did, doing, a, an, the, and, but, if, or, because, as, until, while, of, at, by, for, with, about, against, between, into, through, during, before, after, above, below, to, from, up, down, in, out, on, off, over, under, again, further, then, once, here, there, when, where, why, how, all, any, both, each, few, more, most, other, some, such, no, nor, not, only, own, same, so, than, too, very, s, t, can, will, just, don, don't, should, should've, now, d, ll, m, o, re, ve, y, ain, aren, aren't, couldn, couldn't, didn, didn't, doesn, doesn't, hadn, hadn't, hasn, hasn't, haven, haven't, isn, isn't, ma, mightn, mightn't, mustn, mustn't, needn, needn't, shan, shan't, shouldn, shouldn't, wasn, wasn't, weren, weren't, won, won't, wouldn, wouldn't"

```
[]: #stopwords detection
     set_stopwords = set(stopwords.words("english"))
     def clean txt(text): # define the fuction with tokenization/string cleaning for
      →all datasets
         text = re.sub(r"[^A-Za-z,!?]", " ", text)
         text = re.sub(r'\backslash[[^{?}]]*\backslash]', "", text)
         text = re.sub(r"\s", "", text)
         text = re.sub(r"\'t", "", text)
         text = re.sub(r"\'re", "",text)
         text = re.sub(r"\'d", "", text)
         text = re.sub(r"\'ll", " ", text)
         text = re.sub(r",", " ", text)
         text = re.sub(r"\(", "", text)
         text = re.sub(r"\)", " ", text)
         text = re.sub(r"\'", " ", text)
         text = re.sub(r"aa", "", text)
         text = re.sub(r"zz", "", text)
         text = re.sub(r''[0-9]'', '', text)
         text = text.lower()
         text = ' '.join(word for word in text.split() if word not in set_stopwords)
```

```
return text
     SD_DATA['headline'] = SD_DATA['headline'].apply(clean_txt)
[]: SD_DATA.head(10)
                                                                headline
[]:
        is_sarcastic
     0
                   1 thirtysomething scientists unveil doomsday clo...
     1
                   0 dem rep totally nails congress falling short g...
     2
                              eat veggies deliciously different recipes
                           inclement weather prevents liar getting work
     3
     4
                   1 mother comes pretty close using word streaming...
     5
                                                       white inheritance
                   0
     6
                   0
                                             ways file taxes less stress
     7
                   1 richard branson global warming donation nearly...
```

1 shadow government getting large meet marriott ...

lots parents know scenario

[]: 205988

8 9

```
[]: #checking emojies in the data
     import pandas as pd
     import re
     # Example DataFrame (replace this with your actual DataFrame)
     # SD_DATA = pd.read_csv("your_data.csv")
     def has_emoji(text):
         # Regular expression pattern to match emojis and emoticons
         emoji_pattern = re.compile("["
                                    u"\U0001F600-\U0001F64F" # Emoticons
                                    u"\U0001F300-\U0001F5FF"
                                                              # Symbols & Pictographs
                                    u"\U0001F680-\U0001F6FF"
                                                              # Transport & Map
      ⇔symbols
                                    u"\U0001F1E0-\U0001F1FF"
                                                              # Flags (iOS)
                                    u"\U00002500-\U00002BEF"
                                                              # Chinese/Japanese/
      →Korean characters
                                    u"\U00002702-\U000027B0"
```

```
u"\U00002702-\U000027B0"
                               u"\U000024C2-\U0001F251"
                               u"\U0001f926-\U0001f937"
                               u"\U00010000-\U0010ffff"
                               u"\u2640-\u2642"
                               u"\u2600-\u2B55"
                               u"\u200d"
                               u"\u23cf"
                               u"\u23e9"
                               u"\u231a"
                               u"\ufeOf" # Dingbats
                               u"\u3030"
                               "]+", flags=re.UNICODE)
    # Check if emoji pattern matches any part of the text
   return bool(emoji_pattern.search(text))
# Apply the function to the 'headline' column of SD_DATA
emoji_present = SD_DATA['headline'].apply(has_emoji)
# Check if any emojis or emoticons are present in the data
if emoji_present.any():
   print("Emojis or emoticons are present in the data.")
else:
   print("No emojis or emoticons found in the data.")
```

No emojis or emoticons found in the data.

```
[]: from collections import Counter
    cnt = Counter()
    for text in SD_DATA["headline"].values:
        for word in text.split():
            cnt[word] += 1

cnt.most_common(10)
```

```
[]: #checking most common words
     from collections import Counter
     # Assuming you have a 'headline' column in your DataFrame
     headlines = SD_DATA['headline']
     # Initialize a Counter to count the frequency of words
     word_counts = Counter()
     # Iterate through each headline to count words
     for headline in headlines:
         word_counts.update(headline.split())
     # Determine the most common words to remove
     most_common_words = set(word for word, count in word_counts.most_common(10))
      Adjust 10 to the number of most frequent words you want to remove
     # Function to remove the most frequent words from a headline
     def remove_most_common_words(headline):
         return ' '.join(word for word in headline.split() if word not in_
     →most_common_words)
     # Apply the function to each headline in the DataFrame
     SD_DATA['headline'] = SD_DATA['headline'].apply(remove_most_common_words)
     # Display the modified DataFrame
     print(SD_DATA.head(10))
       is_sarcastic
                                                               headline
    0
                  1 thirtysomething scientists unveil doomsday clo...
                  0 dem rep totally nails congress falling short g...
    1
    2
                             eat veggies deliciously different recipes
                  0
    3
                  1
                          inclement weather prevents liar getting work
                  1 mother comes pretty close using word streaming...
    4
    5
                                                      white inheritance
    6
                  0
                                           ways file taxes less stress
    7
                  1 richard branson global warming donation nearly...
    8
                  1
                    shadow government getting large meet marriott ...
    9
                  0
                                             lots parents know scenario
[]:
[]: #finding the rarer words
     from collections import Counter
     # Assuming you have a 'headline' column in your DataFrame
     headlines = SD_DATA['headline']
```

```
# Initialize a Counter to count the frequency of words
word_counts = Counter()
# Iterate through each headline to count words
for headline in headlines:
   word counts.update(headline.split())
# Determine the rarest words to remove
n rarer words = 10
rarer_words = set([word for word, count in word_counts.most_common()[:
 →-n_rarer_words-1:-1]])
# Function to remove the rarer words from a headline
def remove_rarer_words(headline):
   return ' '.join(word for word in headline.split() if word not in_
 →rarer words)
# Apply the function to each headline in the DataFrame
SD_DATA['headline'] = SD_DATA['headline'].apply(remove_rarer_words)
# Display the modified DataFrame
print(SD_DATA.head(10))
```

```
headline
   is_sarcastic
              1 thirtysomething scientists unveil doomsday clo...
0
              0 dem rep totally nails congress falling short g...
1
2
              0
                         eat veggies deliciously different recipes
                      inclement weather prevents liar getting work
3
              1
4
              1 mother comes pretty close using word streaming...
5
              0
                                                  white inheritance
6
                                        ways file taxes less stress
7
              1 richard branson global warming donation nearly...
              1 shadow government getting large meet marriott ...
8
                                         lots parents know scenario
```

```
[]: #checking emoticons present in the data or not
EMOTICONS = {
    u":-\)":"Happy face or smiley",
    u":-\]":"Happy face or smiley",
    u":-\]":"Happy face or smiley",
    u":-3":"Happy face smiley",
    u":3":"Happy face smiley",
    u":->":"Happy face smiley",
    u":->":"Happy face smiley",
    u":>":"Happy face smiley",
    u":>":"Happy face smiley",
```

```
u":o\)":"Happy face smiley",
u":-\}": "Happy face smiley".
u":\}":"Happy face smiley".
u":-\)":"Happy face smiley",
u":c\)":"Happy face smiley",
u":\^\)":"Happy face smiley",
u"=\]":"Happy face smiley",
u"=\)":"Happy face smiley",
u":-D":"Laughing, big grin or laugh with glasses",
u":D": "Laughing, big grin or laugh with glasses",
u"8-D": "Laughing, big grin or laugh with glasses",
u"8D": "Laughing, big grin or laugh with glasses",
u"X-D": "Laughing, big grin or laugh with glasses",
u"XD": "Laughing, big grin or laugh with glasses",
u"=D": "Laughing, big grin or laugh with glasses",
u"=3": "Laughing, big grin or laugh with glasses",
u"B\^D": "Laughing, big grin or laugh with glasses",
u":-\)\)":"Very happy",
u":-\(":"Frown, sad, andry or pouting",
u":-\(":"Frown, sad, andry or pouting",
u":\(":"Frown, sad, andry or pouting",
u":-c": "Frown, sad, andry or pouting",
u":c":"Frown, sad, andry or pouting",
u":-<":"Frown, sad, andry or pouting",
u":<":"Frown, sad, andry or pouting",
u":-\setminus[":"Frown, sad, andry or pouting",
u":\[":"Frown, sad, andry or pouting",
u":-\|\|":"Frown, sad, andry or pouting",
u">:\[":"Frown, sad, andry or pouting",
u":\{":"Frown, sad, andry or pouting",
u":@":"Frown, sad, andry or pouting",
u">:\(":"Frown, sad, andry or pouting",
u":'-\(":"Crving".
u":'\(":"Crying",
u":'-\)":"Tears of happiness",
u":'\)":"Tears of happiness",
u"D-':":"Horror",
u"D:<":"Disgust",
u"D:":"Sadness".
u"D8": "Great dismay",
u"D;":"Great dismay",
u"D=":"Great dismay",
u"DX": "Great dismay",
u":-0":"Surprise",
u":0":"Surprise",
u":-o":"Surprise",
u":o":"Surprise",
```

```
u":-0": "Shock",
u"8-0":"Yawn",
u">:0":"Yawn".
u":-\*":"Kiss",
u":\*":"Kiss".
u":X":"Kiss",
u''; - \)'': "Wink or smirk",
u";\)":"Wink or smirk",
u'' \times - )'': "Wink or smirk",
u"\*\)":"Wink or smirk",
u":-\]":"Wink or smirk".
u";\]":"Wink or smirk",
u"; \^)": "Wink or smirk",
u":-,":"Wink or smirk",
u";D":"Wink or smirk",
u":-P":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u":P":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u"X-P": "Tongue sticking out, cheeky, playful or blowing a raspberry",
u"XP": "Tongue sticking out, cheeky, playful or blowing a raspberry",
u":-P":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u":P":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u":b":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u"d:":"Tongue sticking out, cheeky, playful or blowing a raspberry",
u"=p": "Tongue sticking out, cheeky, playful or blowing a raspberry",
u">:P": "Tongue sticking out, cheeky, playful or blowing a raspberry",
u":-/": "Skeptical, annoyed, undecided, uneasy or hesitant",
u":/": "Skeptical, annoyed, undecided, uneasy or hesitant",
u":-[.]": "Skeptical, annoyed, undecided, uneasy or hesitant",
u">:[(\\\)]":"Skeptical, annoyed, undecided, uneasy or hesitant",
u">:/": "Skeptical, annoyed, undecided, uneasy or hesitant",
u":[(\\\)]":"Skeptical, annoyed, undecided, uneasy or hesitant",
u"=/": "Skeptical, annoyed, undecided, uneasy or hesitant",
u"=[(\\\)]":"Skeptical, annoyed, undecided, uneasy or hesitant",
u":L": "Skeptical, annoyed, undecided, uneasy or hesitant",
u"=L": "Skeptical, annoyed, undecided, uneasy or hesitant",
u":S": "Skeptical, annoyed, undecided, uneasy or hesitant",
u":-\|":"Straight face",
u":\|":"Straight face",
u":$":"Embarrassed or blushing",
u":-x": "Sealed lips or wearing braces or tongue-tied",
u":x": "Sealed lips or wearing braces or tongue-tied",
u":-#": "Sealed lips or wearing braces or tongue-tied",
u":#": "Sealed lips or wearing braces or tongue-tied",
u":-&":"Sealed lips or wearing braces or tongue-tied",
u":&":"Sealed lips or wearing braces or tongue-tied",
u"O:-\)": "Angel, saint or innocent",
u"0:\)":"Angel, saint or innocent",
```

```
u"0:-3": "Angel, saint or innocent",
  u"0:3": "Angel, saint or innocent",
  u"0:-\)": "Angel, saint or innocent",
  u"0:\)":"Angel, saint or innocent",
  u":-b":"Tongue sticking out, cheeky, playful or blowing a raspberry",
  u"0;\^\)":"Angel, saint or innocent",
  u">:-\)":"Evil or devilish",
  u">:\)":"Evil or devilish",
  u"\setminus\}:-\setminus)":"Evil or devilish",
  u"\}:\)":"Evil or devilish",
  u"3:-\)":"Evil or devilish".
  u"3:\)":"Evil or devilish",
  u">;\)":"Evil or devilish",
  u"\|;-\)":"Cool",
  u"\|-0":"Bored".
  u":-J":"Tongue-in-cheek",
  u"#-\)":"Party all night",
  u"%-\)":"Drunk or confused",
  u"%\)":"Drunk or confused",
  u":-###..":"Being sick",
  u":###..":"Being sick",
  u"<:-\|":"Dump",
  u"\(>_<\)":"Troubled",
  u"\(> <\)>":"Troubled",
  u"\setminus(';'\setminus)":"Baby",
  u"\(\^\^>``":"Nervous or Embarrassed or Troubled or Shy or Sweat drop",
  u''(\hat{\ })'': "Nervous or Embarrassed or Troubled or Shy or Sweat drop",
  u"\(- -;\)": "Nervous or Embarrassed or Troubled or Shy or Sweat drop",
  u"(\sim_~;\) (\.;\)":"Nervous or Embarrassed or Troubled or Shy or Sweat_\(\)
⇔drop",
  u'' (-_-)zzz'': "Sleeping",
  u'' ( \ - ) ": "Wink",
  u'' ((+_+)) : "Confused",
  u'' (\to +) ": "Confused",
  u'' (o | o ) ": "Ultraman",
  u"\^_\^":"Joyful",
  u"\setminus(^_\^)/":"Joyful",
  u"((^0)^) ":"Joyful",
  u"\(\^o\^\) ":"Joyful",
  u"\setminus(_{\_})": "Kowtow as a sign of respect, or dogeza for apology",
  u'' \setminus (\cdot, \cdot)'': "Kowtow as a sign of respect, or dogeza for apology",
  u"<(__\)>":"Kowtow as a sign of respect, or dogeza for apology",
  u"<m\(__\)m>":"Kowtow as a sign of respect, or dogeza for apology",
  u"m\(_\)m":"Kowtow as a sign of respect, or dogeza for apology",
  u"m\(__\)m":"Kowtow as a sign of respect, or dogeza for apology",
  u"\setminus('_')":"Sad or Crying",
  u"\setminus(/_;\setminus)":"Sad or Crying",
```

```
u"\setminus(T_T\setminus)\setminus(;_;\setminus)":"Sad or Crying",
u"\setminus(;_;":"Sad of Crying",
u''(;::)'':"Sad or Crying",
u"\setminus(;0;\setminus)":"Sad or Crying",
u"\setminus(:\_;\setminus)":"Sad or Crying",
u"\(ToT\)":"Sad or Crying",
u"; ;":"Sad or Crying",
u";-;":"Sad or Crying",
u";n;":"Sad or Crying",
u";;":"Sad or Crying",
u"Q\.Q": "Sad or Crying",
u"T\.T": "Sad or Crying",
u"QQ": "Sad or Crying",
u"Q_Q": "Sad or Crying",
u'' \setminus (- \cdot \cdot - \cdot)'' : "Shame",
u"\setminus(-_-\setminus)":"Shame",
u"\setminus(\ \ )": "Shame",
u"\( _\)":"Shame",
u"\setminus(=_=\setminus)":"Tired",
u'' \setminus (= \ \ \ \ \ )'' : "cat",
u'' (= (^\cdot \cdot )^- : "cat",
u"= \^=
                  ":"cat",
u"\setminus(\setminus.\setminus.)":"Looking down",
u''\setminus(\.\.\)'':"Looking down",
u"\^m\^":"Giggling with hand covering mouth",
u"\setminus(\setminus ?":"Confusion".
u"\setminus(?_?\setminus)":"Confusion",
u">\^ \^<":"Normal Laugh",
u"<\^!\^>":"Normal Laugh",
u"\^/\^":"Normal Laugh",
u"\\*\^_\^\* ":"Normal Laugh",
u'' ( ^< ^) ( ^-. ^-) ":"Normal Laugh",
u"\(^\^\)":"Normal Laugh",
u"\setminus(\^\.\^)":"Normal Laugh",
u"\(\^_\^\.\)":"Normal Laugh",
u"\setminus(^_\)":"Normal Laugh",
u"\(\^\^\)":"Normal Laugh",
u"\(\^J\^\)":"Normal Laugh",
u''(\*\^\.\^\*\)'':"Normal Laugh",
u'' ( \ ":"Normal Laugh",
u"\(#\^\.\^#\)":"Normal Laugh",
u'' \ \": "Waving",
u'' \setminus (; ; \setminus) / \sim \sim ": "Waving",
u"((^\.\^\)/~~~":"Waving",
u'' (--)/\sim ((\\cdot \cdot)/\sim ":"Waving",
u'' \setminus (T_T)/\sim ":"Waving",
u"\setminus(ToT\setminus)/\sim\sim":"Waving",
```

```
u''(\*\^0\^*\)'': "Excited",
         u"\(\*_\*\)":"Amazed",
         u"\setminus(\*_\*;":"Amazed",
         u"\(+_+\) \(@_@\)":"Amazed",
         u"\(\*\^\)v":"Laughing,Cheerful",
         u"\setminus(^-\)v":"Laughing,Cheerful",
         u'' ((d[-]b))': "Headphones, Listening to music",
         u'\(-"-\)':"Worried",
         u"\(;\)":"Worried",
         u"\setminus(^0_0^)":"Eyeglasses",
         u"\(\ \\)":"Happy",
         u"\(\ \\)":"Happy",
         u''(\^\)o(\^\)'': "Happy",
         u"\(\^0\^\)":"Happy",
         u"\(\^o\^\)":"Happy",
         u"\)\^o\^\(":"Happy",
         u":0 o_0":"Surprised",
         u"o_0":"Surprised",
         u"o\.0":"Surpised",
         u"\setminus(o\setminus.o\setminus)":"Surprised",
         u"o0": "Surprised",
         u"\setminus(\times m\setminus)":"Dissatisfied",
         u"\setminus(`A`\setminus)":"Snubbed or Deflated"
     }
[]: def remove_emoticons(headline):
         emoticon_pattern = re.compile(u'(' + u'|'.join(k for k in EMOTICONS) + u')')
         return emoticon_pattern.sub(r'', headline)
[]: def remove urls(headline):
         url_pattern = re.compile(r'https?://\S+|www\.\S+')
         return url_pattern.sub(r'', headline)
[]: #removing html tags
     def remove_html(headline):
         html_pattern = re.compile('<.*?>')
         return html_pattern.sub(r'', headline)
     headline = """<div>
     <h1> H2O</h1>
      AutoML
     <a href="https://www.h2o.ai/products/h2o-driverless-ai/"> Driverless AI</a>
     </div>"""
     print(remove_html(headline))
```

H2O AutoML Driverless AI

[]: SD_DATA

[]:	is_sarcastic	headline
0	1	thirtysomething scientists unveil doomsday clo
1	0	dem rep totally nails congress falling short g
2	0	eat veggies deliciously different recipes
3	1	inclement weather prevents liar getting work
4	1	mother comes pretty close using word streaming
•••	•••	
28614	1	jews celebrate something
28615	1	internal affairs investigator disappointed con
28616	0	beautiful acceptance speech week came queer ko
28617	1	mars probe destroyed spielberg gates space palace
28618	1	dad clarifies food stop

[28619 rows x 2 columns]

[]: #short words

```
[]: chat_words_str = """
     AFAIK=As Far As I Know
     AFK=Away From Keyboard
     ASAP=As Soon As Possible
     ATK=At The Keyboard
     ATM=At The Moment
     A3=Anytime, Anywhere, Anyplace
     BAK=Back At Keyboard
     BBL=Be Back Later
     BBS=Be Back Soon
     BFN=Bye For Now
     B4N=Bye For Now
     BRB=Be Right Back
     BRT=Be Right There
     BTW=By The Way
     B4=Before
     B4N=Bye For Now
     CU=See You
     CUL8R=See You Later
     CYA=See You
     FAQ=Frequently Asked Questions
     FC=Fingers Crossed
     FWIW=For What It's Worth
```

```
FYI=For Your Information
GAL=Get A Life
GG=Good Game
GN=Good Night
GMTA=Great Minds Think Alike
GR8=Great!
G9=Genius
IC=I See
ICQ=I Seek you (also a chat program)
ILU=ILU: I Love You
IMHO=In My Honest/Humble Opinion
IMO=In My Opinion
IOW=In Other Words
IRL=In Real Life
KISS=Keep It Simple, Stupid
LDR=Long Distance Relationship
LMAO=Laugh My A.. Off
LOL=Laughing Out Loud
LTNS=Long Time No See
L8R=Later
MTE=My Thoughts Exactly
M8=Mate
NRN=No Reply Necessary
OIC=Oh I See
PITA=Pain In The A..
PRT=Party
PRW=Parents Are Watching
ROFL=Rolling On The Floor Laughing
ROFLOL=Rolling On The Floor Laughing Out Loud
ROTFLMAO=Rolling On The Floor Laughing My A.. Off
SK8=Skate
STATS=Your sex and age
ASL=Age, Sex, Location
THX=Thank You
TTFN=Ta-Ta For Now!
TTYL=Talk To You Later
U=You
U2=You Too
U4E=Yours For Ever
WB=Welcome Back
WTF=What The F...
WTG=Way To Go!
WUF=Where Are You From?
W8=Wait...
7K=Sick:-D Laugher
```

```
[ ]: chat_words_map_dict = {}
     chat_words_list = []
     for line in chat_words_str.split("\n"):
         if line != "":
             cw = line.split("=")[0]
             cw_expanded = line.split("=")[1]
             chat_words_list.append(cw)
             chat_words_map_dict[cw] = cw_expanded
     chat_words_list = set(chat_words_list)
     def chat words conversion(text):
         new_text = []
         for w in text.split():
             if w.upper() in chat_words_list:
                 new_text.append(chat_words_map_dict[w.upper()])
             else:
                 new_text.append(w)
         return " ".join(new_text)
[]: chat_words_conversion("one minute BRB")
[]: 'one minute Be Right Back'
[]:
[]: | #word tokenization
     from nltk.tokenize import word_tokenize
     # Define a function to tokenize the text
     def tokenize_text(headline):
       tokens = word_tokenize(headline)
       return tokens
[]: | # Apply the tokenize_text function to the "headline" column
     SD_DATA["headline"] = SD_DATA["headline"].apply(tokenize_text)
[]: SD_DATA.head(10)
[]:
        is_sarcastic
                                                                headline
                   1 [thirtysomething, scientists, unveil, doomsday...
     0
     1
                   0 [dem, rep, totally, nails, congress, falling, ...
     2
                        [eat, veggies, deliciously, different, recipes]
                   1 [inclement, weather, prevents, liar, getting, ...
     3
     4
                   1 [mother, comes, pretty, close, using, word, st...
     5
                                                    [white, inheritance]
                                       [ways, file, taxes, less, stress]
     6
                   0
     7
                   1 [richard, branson, global, warming, donation, ...
     8
                      [shadow, government, getting, large, meet, mar...
```

```
from wordcloud import WordCloud
import matplotlib.pyplot as plt

# Create a word cloud object
wc = WordCloud(background_color="black", max_words=100)

# Generate the word cloud
wc.generate(str(SD_DATA["headline"]))

# Display the word cloud
plt.imshow(wc, interpolation="bilinear")
plt.axis("off")
plt.show()
```



```
[]: # Create two separate DataFrames for 0 and 1 values
    df_0 = SD_DATA[SD_DATA['is_sarcastic'] == 0]
    df_1 = SD_DATA[SD_DATA['is_sarcastic'] == 1]

# Create a word cloud for each DataFrame
    wc_0 = WordCloud(background_color="black", max_words=100)
    wc_0.generate(str(df_0["headline"]))

wc_1 = WordCloud(background_color="black", max_words=100)
    wc_1.generate(str(df_1["headline"]))

# Display the word clouds
```

```
plt.figure(figsize=(10, 5))

plt.subplot(1, 2, 1)
plt.imshow(wc_0, interpolation="bilinear")
plt.axis("off")
plt.title("Word Cloud for Label 0")

plt.subplot(1, 2, 2)
plt.imshow(wc_1, interpolation="bilinear")
plt.axis("off")
plt.title("Word Cloud for Label 1")

plt.show()
```

Word Cloud for Label 0



2

0

Word Cloud for Label 1



```
[]: # One-hot encoding for categorical features
df_encoded = pd.get_dummies(SD_DATA, columns=['is_sarcastic'])

# Print the encoded DataFrame
print(df_encoded.head(10))
```

```
headline is_sarcastic_0 \
   [thirtysomething, scientists, unveil, doomsday...
                                                                    0
   [dem, rep, totally, nails, congress, falling, ...
1
                                                                    1
2
     [eat, veggies, deliciously, different, recipes]
                                                                      1
   [inclement, weather, prevents, liar, getting, ...
                                                                    0
   [mother, comes, pretty, close, using, word, st...
                                                                    0
4
5
                                 [white, inheritance]
                                                                      1
6
                    [ways, file, taxes, less, stress]
                                                                      1
   [richard, branson, global, warming, donation, ...
                                                                    0
7
8
   [shadow, government, getting, large, meet, mar...
                                                                    0
9
                      [lots, parents, know, scenario]
                                                                      1
   is_sarcastic_1
0
                0
1
```

```
3
                    1
    4
                    1
    5
                    0
    6
                    0
    7
                    1
    8
                    1
    9
                    0
[]: from sklearn.feature_extraction.text import CountVectorizer
[]: vectorizer = CountVectorizer(ngram_range=(2, 2))
[]: X = vectorizer.fit_transform(headlines)
[ ]: vectorizer.get_feature_names_out()
[]: array(['ab exercises', 'abandon everest', 'abandon overly', ...,
            'zuckerberg touts', 'zuckerberg watching', 'zuckerberg wishes'],
           dtype=object)
[]:
[]:
[]: Text = SD_DATA['headline'].values
     op_labels = SD_DATA['is_sarcastic'].values
[]: def _preprocess(doc, accent_function, lower):
         Preprocess a document by joining words into a single string and applying
         lowercasing and accent stripping.
         doc = ' '.join(doc)
         if lower:
             doc = doc.lower()
         if accent_function is not None:
             doc = accent_function(doc)
         return doc
[]: print(type(Text))
    <class 'numpy.ndarray'>
[]: import spacy
[]: nlp = spacy.load("en_core_web_sm")
```

```
[]: Text = [" ".join([token.lemma_ for token in nlp(str(text))]) for text in Text]
[]:
[]: from sklearn.feature_extraction.text import TfidfVectorizer
    multinomial NB
[]: wordVectorizer = TfidfVectorizer(max_features= 1500)
     wordVectorizer.fit(Text)
     features = wordVectorizer.transform(Text)
[]: feature_names = wordVectorizer.get_feature_names_out()
[]: feature_names
[]: array(['abandon', 'abortion', 'absolutely', ..., 'young', 'youth',
            'zuckerberg'], dtype=object)
[]: from sklearn.model selection import train test split
     from sklearn.preprocessing import LabelEncoder
     from sklearn.naive_bayes import MultinomialNB
     from sklearn import metrics
     from sklearn.metrics import accuracy_score
     from sklearn.feature_extraction.text import TfidfVectorizer
[]: xTrain, xTest, yTrain, yTest = train_test_split(features, op_labels, test_size=_
      →0.25)
[]: mnb = MultinomialNB()
[]: mnb.fit(xTrain, yTrain)
[]: MultinomialNB()
[]: mnb.score(xTrain, yTrain)
[]: 0.7614610510622437
[]: predicted = mnb.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                  precision
                             recall f1-score
                                                  support
               0
                       0.74
                                 0.77
                                           0.75
                                                     3757
                       0.73
                                 0.69
               1
                                           0.71
                                                     3398
```

```
macro avg
                       0.73
                                 0.73
                                            0.73
                                                      7155
    weighted avg
                       0.73
                                 0.73
                                           0.73
                                                      7155
[]: #logistic regression
[]: from sklearn.linear_model import LogisticRegression
     lr = LogisticRegression()
     lr.fit(xTrain, yTrain)
[]: LogisticRegression()
[]: lr.score(xTrain, yTrain)
[]: 0.7735277674245248
[]: predicted = lr.predict(xTest)
     print(metrics.classification_report(yTest, predicted))
                  precision
                               recall f1-score
                                                  support
               0
                       0.74
                                 0.77
                                           0.75
                                                      3757
               1
                       0.73
                                 0.71
                                           0.72
                                                      3398
                                                     7155
                                           0.74
        accuracy
       macro avg
                       0.74
                                 0.74
                                           0.74
                                                      7155
    weighted avg
                       0.74
                                 0.74
                                           0.74
                                                     7155
[]: #Decision tree
[]: from sklearn import tree
     dt = tree.DecisionTreeClassifier()
     dt.fit(xTrain, yTrain)
[ ]: DecisionTreeClassifier()
[]: dt.score(xTrain, yTrain)
[]: 0.9849981364144614
[]: predicted = dt.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
```

0.73

accuracy

```
0
                       0.69
                                  0.69
                                            0.69
                                                      3757
               1
                       0.66
                                  0.66
                                            0.66
                                                      3398
        accuracy
                                            0.67
                                                      7155
       macro avg
                       0.67
                                  0.67
                                            0.67
                                                      7155
    weighted avg
                                  0.67
                       0.67
                                            0.67
                                                      7155
[]: #Random forest
[]: from sklearn.ensemble import RandomForestClassifier
     rf = RandomForestClassifier()
     rf.fit(xTrain, yTrain)
[ ]: RandomForestClassifier()
[]: rf.score(xTrain, yTrain)
[]: 0.9849981364144614
[]: predicted = rf.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                  precision
                               recall f1-score
                                                   support
                       0.73
                                  0.75
               0
                                            0.74
                                                      3757
               1
                       0.72
                                  0.70
                                            0.71
                                                      3398
        accuracy
                                            0.72
                                                      7155
                                            0.72
                                                      7155
       macro avg
                       0.72
                                  0.72
    weighted avg
                       0.72
                                 0.72
                                            0.72
                                                      7155
[1]: #svm
[]: from sklearn.svm import LinearSVC
     lsvm = LinearSVC()
     lsvm.fit(xTrain, yTrain)
[]: LinearSVC()
[]: lsvm.score(xTrain, yTrain)
```

recall f1-score

support

precision

```
[]: 0.7776742452478569
[]: predicted = lsvm.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                  precision
                               recall f1-score
                                                   support
               0
                       0.75
                                  0.75
                                            0.75
                                                      3757
               1
                       0.72
                                  0.72
                                            0.72
                                                      3398
        accuracy
                                            0.74
                                                      7155
       macro avg
                       0.74
                                  0.74
                                            0.74
                                                      7155
                                  0.74
                                            0.74
    weighted avg
                       0.74
                                                      7155
[]: #KNeighbors
[]: from sklearn.neighbors import KNeighborsClassifier
     knn = KNeighborsClassifier()
     knn.fit(xTrain, yTrain)
[]: KNeighborsClassifier()
[]: knn.score(xTrain, yTrain)
[]: 0.8442508386134924
[]: predicted = knn.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                  precision
                               recall f1-score
                                                   support
               0
                        0.58
                                  0.76
                                            0.66
                                                      3757
               1
                       0.59
                                  0.39
                                            0.47
                                                      3398
                                            0.58
                                                      7155
        accuracy
                                            0.56
                                                      7155
       macro avg
                       0.59
                                  0.57
    weighted avg
                       0.59
                                  0.58
                                            0.57
                                                      7155
[]: #SGDC Classifier
[]: from sklearn.linear_model import SGDClassifier
```

```
sgd = SGDClassifier()
     sgd.fit(xTrain, yTrain)
[]: SGDClassifier()
[]: sgd.score(xTrain, yTrain)
[]: 0.7733414088706672
[]: predicted = sgd.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                               recall f1-score
                                                   support
                  precision
               0
                       0.74
                                 0.76
                                            0.75
                                                      3757
                       0.73
                                  0.71
                                            0.72
                                                      3398
               1
                                            0.74
                                                      7155
        accuracy
                                                      7155
       macro avg
                       0.74
                                 0.73
                                            0.74
    weighted avg
                       0.74
                                 0.74
                                            0.74
                                                      7155
[]: #voting classifier
[]: from sklearn.ensemble import VotingClassifier
     voting = VotingClassifier(estimators = [('lr', lr),
                                             ('dt', dt),
                                             ('knn', knn),
                                             ('sgd', sgd),
                                             ('mnb', mnb),
                                             ('rf', rf),
                                             ('lsvm', lsvm)],
                              voting = 'hard')
     voting.fit(xTrain, yTrain)
[]: VotingClassifier(estimators=[('lr', LogisticRegression()),
                                  ('dt', DecisionTreeClassifier()),
                                  ('knn', KNeighborsClassifier()),
                                  ('sgd', SGDClassifier()), ('mnb', MultinomialNB()),
                                  ('rf', RandomForestClassifier()),
                                  ('lsvm', LinearSVC())])
[]: voting.score(xTrain, yTrain)
[]: 0.814340290719344
```

```
[]: predicted = voting.predict(xTest)
[]: print(metrics.classification_report(yTest, predicted))
                  precision
                               recall f1-score
                                                   support
               0
                                 0.77
                       0.74
                                           0.76
                                                      3757
               1
                       0.74
                                 0.71
                                            0.72
                                                      3398
                                           0.74
                                                      7155
        accuracy
       macro avg
                       0.74
                                 0.74
                                            0.74
                                                     7155
    weighted avg
                       0.74
                                 0.74
                                           0.74
                                                     7155
[]: for clf in (mnb, lsvm, lr, knn, sgd, dt, rf):
         clf.fit(xTrain, yTrain)
         yPred = clf.predict(xTest)
         print(clf.__class__.__name__, accuracy_score(yTest, yPred))
    MultinomialNB 0.7334730957372467
    LinearSVC 0.735988819007687
    LogisticRegression 0.7385045422781272
    KNeighborsClassifier 0.5828092243186582
    SGDClassifier 0.735988819007687
    DecisionTreeClassifier 0.6788259958071279
    RandomForestClassifier 0.7228511530398323
[]:
[]:
[]:
```