

Covid-19 Vaccine Sentiment Analysis

Goal: To examine tweets recorded about the Covid-19 vaccine to analyze the sentiments of people for the vaccine.

```
In [12]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

In [2]: import warnings
warnings.filterwarnings('ignore')

In [186]: df = pd.read_csv("vaccination_tweets.csv")

Out[4]:
```

		id	user_name	user_location	user_description	user_created	user_followers	user_friends	user_favorites	user_verified
0	134053911971516416	Rachel Roh	La Crescenta-Montrose, CA	Aggregator of Asian American news; scanning d...	2009-04-08 17:52:46	405	1692	3247	F	
1	1338158543359250433	Albert Fong	San Francisco, CA	Marketing dude, tech geek, heavy metal & '80s ...	2009-09-21 15:27:30	834	666	178	F	
2	1337858199140118533	🇪🇺🇺🇦🇵🇸	Your Bed	hell, hydra 🍷👉	2020-06-25 23:30:28	10	88	155	F	
3	1337855739919835717	Charles Adler	Vancouver, BC - Canada	Hosting 'CharlesAdlerTonight' Global News Rad...	2008-09-10 11:28:53	49165	3933	21853	F	
4	1337854064604966912	Citizen News Channel	NaN	Citizen News Channel bringing you an alternat...	2020-04-23 17:58:42	152	580	1473	F	

```
In [187]: df.columns

Out[187]: Index(['id', 'user_name', 'user_location', 'user_description', 'user_created', 'user_followers', 'user_friends', 'user_favorites', 'user_verified', 'date', 'text', 'hashtags', 'source', 'retweets', 'favorites', 'is_retweet', 'dtype='object'])

In [7]: df.dtypes

Out[7]: id                int64
user_name         object
user_location     object
user_description  object
user_created      object
user_followers    int64
user_friends      int64
user_verified     bool
date             object
text             object
hashtags         object
source           object
retweets         int64
favorites         int64
is_retweet       bool
dtype: object

In [188]: df['source'].unique()

Out[188]: array(['Twitter for Android', 'Twitter Web App', 'Twitter for iPhone', 'TweetDeck', 'Buffer', 'Twitter for iPad', 'Twitter Media Studio', 'ThreadReaderApp', 'Instagram', 'SocialFlow', 'Hootsuite Inc.', 'LinkedIn', 'Twitter for Mac', '24liveblog', 'Publer', 'IFTTT', 'Socialbakers', 'Falcon Social Media Management', 'EchoBox', 'Microsoft Power Platform', 'Nonli', 'Sendible', 'nan', 'EastMejo', 'Twitter Media Studio - LiveOut', 'Tweetbot for Mac', 'GT Backend', 'SocialNewsDesk', 'Sprout Social', 'TweetCaster for Android', 'UberSocial for Android', 'BlogSocial APP', 'WordPress.com', 'Paper.li', 'News Users', 'ETRetail.com', 'Tweetbot for iOS', 'Sprinkl Publishing', 'CoSchedule', 'HocalWire Social Share', 'Tumblr', 'loomly', 'The Tweeted Times', 'Fenix 2', 'Imminent News', 'Sprinklr', 'Salesforce - Social Studio', 'Ecoforn', 'Article Tweetbot', 'SEMrush Social Media Tool', 'HashTrends', 'Threader_client', 'Revive Social App', 'Business Upturn', 'Hypefury', 'LaterMedia', 'SocialBe.io v2', 'BizToc News', 'BizToc', 'Social-Pull-Staging'], dtype=object)

In [9]: df['source'].value_counts()

Out[9]: Twitter for Android    58722
Twitter Web App       57219
Twitter for iPhone    49479
cowin_vaccine_app    11679
CowinAlertsBot       11276
...
Btily                 1
Injummie              1
todayvinevada        1
Gravity Forever       1
VitaminRush          1
Name: source, Length: 379, dtype: int64

In [189]: df['source'].value_counts().nlargest(30)

Out[189]: Twitter for iPhone    3579
Twitter Web App      3258
Twitter for Android  2786
TweetDeck            618
Twitter for iPad     272
Instagram            180
Buffer              66
Hootsuite Inc.      22
IFTTT               17
BlogSocial APP      15
LinkedIn            17
Microsoft Power Platform  18
Tweetbot for iOS    12
Nonli              10
Revive Social App   10
dlyr.li            8
WordPress.com       8
Twitter for Mac     7
24liveblog          7
SocialFlow          7
Sprinkl Publishing  7
ETRetail.com        7
Sprout Social       6
Article Tweetbot    6
Salesforce - Social Studio  5
Twitter Media Studio 5
Flying Eze          5
Twitter Media Studio - LiveOut 4
Name: source, dtype: int64

In [110]: plt.figure(figsize=(20,10))
sns.barplot(x='source', y='favorites', data=df)
plt.xticks(rotation=45)
plt.show()
```

Text Analysis of tweets

```
In [111]: import re
import nltk
import nltk.tokenize
import nltk.tokenize
import nltk.tokenize
import nltk.tokenize

In [112]: df['text'].iloc[2]

Out[112]: '#coronavirus #SputnikV #AstraZeneca #PfizerBioNTech #Moderna #Covid_19 Russian vaccine is created to last 2-4 y ears. https://t.co/ieYlCKBr8P'

In [113]: df['text'].apply(nfx.extract_hashtags)

Out[113]: 0      []
1      []
2      []
3      []
4      []
...
11015  []
11016  []
11017  []
11018  []
11019  []
Name: text, Length: 11020, dtype: object

In [114]: df['extracted_hashtags'] = df['text'].apply(nfx.extract_hashtags)

In [115]: df

Out[115]:
```

		id	user_name	user_location	user_description	user_created	user_followers	user_friends	user_favourite
0	134053911971516416	Rachel Roh	La Crescenta-Montrose, CA	Aggregator of Asian American news; scanning d...	2009-04-08 17:52:46	405	1692	324	
1	1338158543359250433	Albert Fong	San Francisco, CA	Marketing dude, tech geek, heavy metal & '80s ...	2009-09-21 15:27:30	834	666	17	
2	1337858199140118533	🇪🇺🇺🇦🇵🇸	Your Bed	hell, hydra 🍷👉	2020-06-25 23:30:28	10	88	15	
3	1337855739919835717	Charles Adler	Vancouver, BC - Canada	Hosting 'CharlesAdlerTonight' Global News Rad...	2008-09-10 11:28:53	49165	3933	2185	
4	1337854064604966912	Citizen News Channel	NaN	Citizen News Channel bringing you an alternat...	2020-04-23 17:58:42	152	580	147	
...	
11015	1460932683630977025	theSun	Petaling Jaya	Breaking news updates from the website of Mala...	2009-07-29 01:59:44	151937	465	59	
11016	1460809305072896516	Sujan	NaN	—Napricorn— it's jus...	2013-01-28 11:46:20	4259	622	15694	
11017	1460581918295031809	theSun	Petaling Jaya	Breaking news updates from the website of Mala...	2009-07-29 01:59:44	151937	465	59	
11018	1460344277951782922	Aleksajs Jackovs	Ireland	#Scientificresearcher in #InformationTechnology...	2012-08-13 13:47:03	1	38	9	
11019	1460208097596225028	theSun	Petaling Jaya	Breaking news updates from the website of Mala...	2009-07-29 01:59:44	151937	465	59	

11020 rows × 17 columns

```
In [116]: #remove hashtags
df['clean_tweet'] = df['text'].apply(nfx.remove_hashtags)

In [117]: df[['text', 'clean_tweet']]

Out[117]:
```

		text	clean_tweet
0	Same folks said daikon paste could treat a cyt...	Same folks said daikon paste could treat a cyt...	
1	While the world has been on the wrong side of ...	While the world has been on the wrong side of ...	
2	#coronavirus #SputnikV #AstraZeneca #PfizerBio...	Russian vaccine is created to last 2-4 years...	
3	Facts are immutable, Senator, even when you're...	Facts are immutable, Senator, even when you're...	
4	Explain to me again why we need a vaccine @Bor...	Explain to me again why we need a vaccine @Bor...	
...
11015	The number of Covid-19 cases today, 17 Novembe...	The number of Covid-19 cases today, 17 Novembe...	
11016	First dose [green checkmark] #PfizerBioNTech	First dose [green checkmark]	
11017	The number of Covid-19 cases today, 16 Novembe...	The number of Covid-19 cases today, 16 Novembe...	
11018	#PfizerBioNTech has developed new pill #Paxlov...	has developed new pill that increased the cha...	
11019	The number of Covid-19 cases today, 15 Novembe...	The number of Covid-19 cases today, 15 Novembe...	

11020 rows × 2 columns

```
In [118]: df['clean_tweet'] = df['clean_tweet'].apply(lambda x:nfx.remove_userhandles(x))

In [119]: #remove multiple white spaces
df['clean_tweet'] = df['clean_tweet'].apply(nfx.remove_multiple_spaces)

In [120]: df['clean_tweet']

Out[120]: 0      Same folks said daikon paste could treat a cyt...
1      While the world has been on the wrong side of ...
2      Russian vaccine is created to last 2-4 years...
3      Facts are immutable, Senator, even when you're...
4      Explain to me again why we need a vaccine http...
...
11015  The number of Covid-19 cases today, 17 Novembe...
11016  First dose [green checkmark] #PfizerBioNTech
11017  The number of Covid-19 cases today, 16 Novembe...
11018  has developed new pill that increased the cha...
11019  The number of Covid-19 cases today, 15 Novembe...
Name: clean_tweet, Length: 11020, dtype: object

In [121]: #remove urls
df['clean_tweet'] = df['clean_tweet'].apply(nfx.remove_urls)

In [122]: #remove punctuations
df['clean_tweet'] = df['clean_tweet'].apply(nfx.remove_puncts)

In [123]: df[['text', 'clean_tweet']]

Out[123]:
```

		text	clean_tweet
0	Same folks said daikon paste could treat a cyt...	Same folks said daikon paste could treat a cyt...	
1	While the world has been on the wrong side of ...	While the world has been on the wrong side of ...	
2	#coronavirus #SputnikV #AstraZeneca #PfizerBio...	Russian vaccine is created to last 2-4 years...	
3	Facts are immutable, Senator, even when you're...	Facts are immutable, Senator, even when you're...	
4	Explain to me again why we need a vaccine @Bor...	Explain to me again why we need a vaccine @Bor...	
...
11015	The number of Covid-19 cases today, 17 Novembe...	The number of Covid19 cases today 17 November...	
11016	First dose [green checkmark] #PfizerBioNTech	First dose [green checkmark]	
11017	The number of Covid-19 cases today, 16 Novembe...	The number of Covid19 cases today 16 November...	
11018	#PfizerBioNTech has developed new pill #Paxlov...	has developed new pill that increased the cha...	
11019	The number of Covid-19 cases today, 15 Novembe...	The number of Covid19 cases today 15 November...	

11020 rows × 2 columns

Sentiment Analysis

```
In [124]: from textblob import TextBlob

In [159]: def get_sentiment(text):
    blob = TextBlob(text)
    sentiment_polarity = blob.sentiment.polarity
    sentiment_subjectivity = blob.sentiment.subjectivity
    if sentiment_polarity > 0:
        sentiment_label = 'Positive'
    elif sentiment_polarity < 0:
        sentiment_label = 'Negative'
    else:
        sentiment_label = 'Neutral'
    result = {'polarity':sentiment_polarity,
             'subjectivity':sentiment_subjectivity,
             'sentiment':sentiment_label}
    return result

In [160]: ex1 = df['clean_tweet'].iloc[0]

In [161]: get_sentiment(ex1)

Out[161]: {'polarity': 0.0, 'subjectivity': 0.125, 'sentiment': 'Neutral'}

In [162]: df['sentiment_results'] = df['clean_tweet'].apply(get_sentiment)

In [210]: pd.json_normalize(df['sentiment_results'])

Out[210]:
```

		polarity	subjectivity	sentiment
0	0.000000	0.125000	Neutral	
1	-0.500000	0.900000	Negative	
2	0.000000	0.033333	Neutral	
3	-0.050000	0.550000	Negative	
4	0.000000	0.000000	Neutral	
...	
11015	0.000000	0.000000	Neutral	
11016	0.250000	0.333333	Positive	
11017	0.000000	0.000000	Neutral	
11018	-0.121212	0.501515	Negative	
11019	0.000000	0.000000	Neutral	

11020 rows × 3 columns

```
In [212]: df

Out[212]:
```

		id	user_name	user_location	user_description	user_created	user_followers	user_friends	user_favourite
0	134053911971516416	Rachel Roh	La Crescenta-Montrose, CA	Aggregator of Asian American news; scanning d...	2009-04-08 17:52:46	405	1692	324	
1	1338158543359250433	Albert Fong	San Francisco, CA	Marketing dude, tech geek, heavy metal & '80s ...	2009-09-21 15:27:30	834	666	17	
2	1337858199140118533	🇪🇺🇺🇦🇵🇸	Your Bed	hell, hydra 🍷👉	2020-06-25 23:30:28	10	88	15	
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...	
11015	1460932683630977025	theSun	Petaling Jaya	Breaking news updates from the website of Mala...	2009-07-29 01:59:44	151937	465	59	
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11018	1460344277951782922	Aleksajs Jackovs	Ireland	#Scientificresearcher in #InformationTechnology...	2012-08-13 13:47:03	1	38	9	
11019	1460208097596225028	theSun	Petaling Jaya	Breaking news updates from the website of Mala...	2009-07-29 01:59:44	151937	465	59	

11020 rows × 22 columns

```
In [166]: df['sentiment'].value_counts()

Out[166]: Neutral    5243
Positive    4638
Negative    1139
Name: sentiment, dtype: int64

In [167]: df['sentiment'].value_counts().plot(kind='bar')

Out[167]: <AxesSubplot>
```

```
In [168]: sns.countplot(df['sentiment'])

Out[168]: <AxesSubplot: xlabel='sentiment', ylabel='count'>
```

```
In [204]: positive_tweet = df[df['sentiment'] == 'Positive']['clean_tweet']
negative_tweet = df[df['sentiment'] == 'Negative']['clean_tweet']
neutral_tweet = df[df['sentiment'] == 'Neutral']['clean_tweet']

In [216]: positive_tweet.head(10)

Out[216]: 5      Does anyone have any helpful advice/guidance fo...
6      There have not been many bright days in 2020 b...
10     While deaths are closing in on the 300000 mark...
11     Best wishes to the US
12     The agency also released new information for h...
15     Trump announces rollout in less than 24 hours...
17     Coronavirus: Iran reports 8201 new cases 221 d...
20     How much did the get paid to approve this all ...
22     Trump announces rollout in less than 24 hours ...
24     Presenting you the top Medical News/Advancemen...
Name: clean_tweet, dtype: object

In [217]: neutral_tweet.head(10)

Out[217]: 0      Same folks said daikon paste could treat a cyt...
1      Russian vaccine is created to last 2-4 years...
2      Explain to me again why we need a vaccine @Bor...
3      Covid vaccine you getting it
4      States will start getting Monday says
13     For all the women and healthcare providers who...
14     Expect 145 sites across all the states to rece...
15     UPAATEO: amp Part Two
21     Anyone wondering why day after approval in the...
23     The US Food and Drug Administration (FDA) has ...
Name: clean_tweet, dtype: object

In [218]: negative_tweet.head(10)

Out[218]: 1      While the world has been on the wrong side of ...
3      Facts are immutable Senator even when youre no...
4      it is a bit sad to claim the fame for success
18     will rake in billions from its expensive but ...
19     The trump administration failed to deliver on ...
26     Wear a mask, wash your hands and remain social...
31     Please don't spread fake news There is no evi...
33     All respect to due process but why not give ap...
39     Check out table 25 of this and other tables t...
50     Tragedy Another life taken by corporate greed ...
Name: clean_tweet, dtype: object

In [224]: positive_tweet_list = positive_tweet.apply(nfx.remove_stopwords).tolist()
negative_tweet_list = negative_tweet.apply(nfx.remove_stopwords).tolist()
neutral_tweet_list = neutral_tweet.apply(nfx.remove_stopwords).tolist()

In [175]: pos_tokens = [token for line in positive_tweet_list for token in line.split()]
neg_tokens = [token for line in negative_tweet_list for token in line.split()]
neut_tokens = [token for line in neutral_tweet_list for token in line.split()]

In [177]: #Get most commonest keywords
from collections import Counter

In [178]: def get_tokens(docx,num=30):
    word_tokens = Counter(docx)
    most_common = word_tokens.most_common(num)
    result = dict(most_common)
    return result

In [179]: get_tokens(pos_tokens)

Out[179]: {'vaccine': 1355,
'dose': 748,
'today': 349,
'Pfizer': 278,
'got': 273,
'COVID19': 233,
'effective': 215,
'amp': 210,
'received': 209,
'vaccines': 203,
'doses': 184,
'shot': 182,
'new': 177,
'second': 177,
'people': 175,
'vaccinated': 174,
'vaccination': 163,
'vaccine': 163,
'jab': 149,
'Thanks': 147,
'2nd': 147,
'got': 139,
'good': 137,
'COVID': 136,
'1': 132,
'Thank': 132,
'day': 118,
'the': 118,
'1st': 107,
'2': 105}

In [180]: most_common_pos_words = get_tokens(pos_tokens)
most_common_neg_words = get_tokens(neg_tokens)
most_common_neut_words = get_tokens(neut_tokens)

In [181]: pos_df = pd.DataFrame(most_common_pos_words.items(), columns=['words', 'scores'])

In [182]: pos_df.head(10)

Out[182]:
```

		words	scores
0	vaccine	1355	
1	dose	748	
2	today	349	
3	Pfizer	278	
4	got	273	
5	COVID19	233	
6	effective	215	
7	amp	210	
8	received	209	
9	vaccines	203	

```
In [183]: plt.figure(figsize=(20,10))
sns.barplot(x='words', y='scores', data=pos_df)
plt.xticks(rotation=45)
plt.show()
```

```
In [184]: neut_df = pd.DataFrame(most_common_neut_words.items(), columns=['words', 'scores'])
neut_df.head(10)

Out[184]:
```

		words	scores
0	vaccine	1147	
1	dose	454	
2	today	313	
3	Pfizer	255	
4	doses	241	
5	COVID19	239	
6	vaccinated	237	
7	COVID19	232	
8	2nd	219	
9	second	202	

```
In [185]: plt.figure(figsize=(20,10))
sns.barplot(x='words', y='scores', data=neut_df)
plt.xticks(rotation=45)
plt.show()
```

```
In [186]: neg_df = pd.DataFrame(most_common_neg_words.items(), columns=['words', 'scores'])
neg_df.head(10)

Out[186]:
```

		words	scores
0	vaccine	1147	
1	dose	454	
2	today	313	
3	Pfizer	255	
4	2nd	62	
5	amp	58	
6	arm	51	
7	people	51	
8	today	46	
9	shot	43	

```
In [152]: # Word Cloud
from wordcloud import WordCloud

In [213]: def plot_wordcloud(docx):
    plt.figure(figsize=(20,10))
    sns.barplot(x='words', y='scores', data=pos_df)
    plt.title('Common Words Among Most Positive Tweets', fontsize=16, fontweight='bold')
    mywordcloud = WordCloud().generate(docx)
    plt.imshow(mywordcloud, interpolation='bilinear')
    plt.axis('off')
    plt.show()

In [214]: pos_docx = ' '.join(pos_tokens)
neg_docx = ' '.join(neg_tokens)
neut_docx = ' '.join(neut_tokens)

In [215]: plot_wordcloud(pos_docx)
plot_wordcloud(neg_docx)
```

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

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Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.

Common Words Among Most Negative Tweets

Word cloud for negative tweets. The most prominent words are 'vaccine', 'dose', 'today', 'Pfizer', 'got', 'COVID19', 'effective', 'amp', 'received', 'vaccines', 'doses', 'shot', 'new', 'second', 'people', 'vaccinated', 'vaccination', 'jab', 'Thanks', '2nd', 'good', 'COVID', 'Thank', 'day', 'the', '1st', '2'.