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Article Sentiment Computation of UK-originated COVID-19 Vaccine Tweets: A Chronological Analysis and News Effect

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Abstract: This study aimed to analyse public sentiments of UK-originated tweets about COVID-19 1 vaccines using six chronological data periods between January and December 2021. The dates are 2 based on six BBC news reports about the most significant developments in the three main vaccines з administered in the UK - Pfizer-BioNTech, Moderna, and Oxford-AstraZeneca. Each data period 4 spans seven days, starting from the news report. The study employed the Bidirectional Encoder 5 Representations from Transformers (BERT) model to analyse the sentiments in the 4,172 extracted 6 tweets. The BERT model adopts the transformer architecture and uses the 'Masked Language Model' and 'Next Sentence Prediction'. The results show that the overall sentiments for all three vaccines 8 were negative across all six periods, with Moderna having the least negative tweets and the highest 9 percentage of positive tweets overall, while AstraZeneca attracted the most negative tweets. However, 10 for all the considered periods, period 3 (23 -29 May 2021) received the least negative and the most 11 positive tweets, following the BBC report - COVID - Pfizer and AstraZeneca jabs work against Indian 12 variant, despite reports of blood clot cases associated with AstraZeneca in the same period. Periods 13 5 to 6, where there was no breaking news relating to COVID Vaccines, had no significant changes. 14 We, therefore, conclude that the BBC News reports on COVID Vaccines significantly impacted public 15 sentiments regarding the COVID-19 Vaccines. 16

Keywords: COVID-19; sentiment Analysis; Tweets; Breaking News; Vaccines

1. Introduction

The novel coronavirus, SARS-CoV-2, behind the coronavirus disease, COVID-19, was 19 first detected in Wuhan, Hubei Province, People's Republic of China, in December 2019 20 and was declared a pandemic by the World Health Organisation (WHO) on 11 March 2020. 21 By 27 October 2021, there has been more than 4.9 million COVID-19-related deaths and 22 245.5 million confirmed COVID-19 cases reported globally [1]. As of 27 October 2021, more 23 than 6.94 billion doses of COVID-19 vaccines have been administered worldwide, which 24 enabled at least 48.9% of the total global population to be at least partly vaccinated, and 25 37.9% fully vaccinated against COVID-19 (Our World in Data 2021:a). While 272 vaccines 26 were in development against COVID-19, 104 vaccines were in clinical testing, and 22 were 27 in use on 29 October 2021 (Milken Institute 2021), it is the Pfizer-BioNTech, Moderna, and 28 Oxford-AstraZeneca COVID-19 vaccines that are of interest to this study owing to their 29 crucial role in the UK's mass immunisation programme [2]. The pandemic disrupted a 30 lot of activities all over the world; schools and education systems were halted and later 31 proceeded to online studies [3,4]. Another worrying devlopment of the pandemic was 32 information overload, some of which were fake news [5], leading to contrasting opinions 33 from the general public [6,7]. 34

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1.1. The COVID-19 Pandemic and Vaccination in the UK

The first officially confirmed COVID-19 case and death in the United Kingdom (UK) 36 were discovered in January and March, respectively. The UK was the first western country 37 to authorise a COVID-19 vaccine for emergency use with the approval by the Medicines and 38 Healthcare products Regulatory Agency (MHRA) on 2 December 2020 and subsequently 39 became the first country to start mass immunisation with the Pfizer-BioNTech COVID-40 19 vaccine on 8 December 2020. The Oxford-AstraZeneca vaccine was approved in the 41 UK by the MHRA on 30 December 2020 and was administered outside of clinical trials 42 on 4 January 2021, whereas the Moderna COVID-19 vaccine was first administered on 43 7 April 2021. Promoting COVID-19 vaccination uptake was necessary for reducing the 44 strain on the National Health Service (NHS), especially since the number of patients in UK 45 hospitals with a confirmed COVID-19 infection had been mainly on the rise. The number 46 of patients in mechanical ventilation (MV) beds in UK hospitals also increased significantly. 47 Furthermore, the number of daily deaths within 28 days of testing positive for COVID-19 48 has increased approximately 20-fold in five months: from 10 deaths on 27 May 2021 to 49 207 deaths on 27 October 2021 in the UK. By 27 October 2021, there has been more than 50 140,000 confirmed deaths within 28 days of testing positive for COVID-19 and 8.8 million 51 confirmed COVID-19 cases reported in the UK. However, as at 27 October 2021, there has 52 also been more than 102.1 million doses of COVID-19 vaccines administered in the UK, 53 with more than 49.7 million recipients of the first COVID-19 vaccine dose [8]. Moreover, the COVID-19 booster vaccination programme was already underway by 27 October 2021 55 for people most at risk from COVID-19 without the third dose of vaccine, thus promoting vaccine uptake continues to be indispensable for keeping people protected against severe 57 illness [9].

1.2. COVID-19 vaccine sentiment on Twitter

Government-imposed public health interventions, such as stay-at-home orders or quar-60 antines, introduced additional challenges for the traditional methods used for gathering 61 data on people's opinions [10]. For instance, utilising surveys for data collection had draw-62 backs even before the COVID-19 pandemic, manifesting in their small and unrepresentative 63 samples or relatively expensive and time-consuming implementation [11,12]. Therefore, 64 researchers like Hussain et al. [13] advocate using social media data to examine public 65 sentiments due to the possibility of acquiring more representative samples with the right 66 spatiotemporal granularity often at a lower cost from social media platforms than surveys. 67 Social media is a suitable data source for this project as around 82% of internet-using adults 68 in the UK have a profile on social media nowadays with roughly consistent representation 69 from all socio-economic groups and genders [14]. One of the largest social networking 70 platforms is Twitter with over 211 million daily active users around the world in the third 71 quarter of 2021 which makes it a valuable data source for conducting sentiment analysis on 72 COVID-19 vaccine-related discourse. Moreover, Twitter users have been thus far willing to 73 express their views relatively openly on this social networking platform due to the sense 74 of anonymity they have generally had [15] and the lack of stringent scientific vetting or 75 editorial curation of tweets on the platform [16]. Consequently, Twitter can be a valuable 76 data source for researching public sentiment towards COVID-19 vaccines. Legitimate news 77 events reported by authoritative news organisations worldwide could potentially influence the sentiment towards COVID-19 vaccines among Twitter users [17]. This project is set to 79 analyse the sentiment of UK-originated tweets and the British Broadcasting Corporation 80 (BBC) continues to be the news organisation with the highest reach among UK-based adults, 81 the BBC can be regarded as the most influential authoritative news organisation in the UK. 82

1.3. The benefit of understanding COVID-19 vaccine sentiments

Sentiment analysis of UK-originated tweets about specific COVID-19 vaccines is one possible avenue for understanding public sentiment towards COVID-19 vaccines in the UK. As sentiment analysis can be conducted retrospectively, this project will focus on the days

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following the publication of certain BBC news reports. These news articles were selected owing to their perceived potential to generate positive or negative sentiment towards the three COVID-19 vaccines used in the UK's vaccination programme. By exploring the general sentiment about COVID-19 vaccines at these times, examining the likely effects of the news reports on the public sentiment towards COVID-19 vaccines, and comparing the sentiment towards different COVID-19 vaccines over time, this project may provide useful insights for the UK authorities.

The contribution of this paper is 4-fold:

- To explore the general sentiment about COVID-19 vaccines in the UK
- To chronologically examine the potential effects of news reports on the public sentiment towards COVID-19 vaccines
- To compare the sentiments toward different COVID-19 vaccines in line with these news reports
- To compare the sentiment towards COVID-19 vaccines over time

Many studies have researched sentiment analysis of the public opinion on COVID-19 101 vaccines, especially on the effect of media on these opinions. However, there are still 102 some research gaps that this study seeks to fill. To the best of our knowledge, no existing 103 study performed a chronological analysis to understand how news reports impact people's 104 sentiments. Particularly, there is a lack of research on understanding public sentiment 105 towards COVID-19 vaccines specifically in the UK context. This study aims to address 106 this gap by analyzing tweets originating from the UK to understand public sentiment 107 towards COVID-19 vaccines over time. Also, there is limited research on the impact of news and events on public sentiment towards vaccines. This study aims to address 109 this gap by chronologically analyzing tweets in relation to specific news and events to 110 understand how they impact public sentiment towards COVID-19 vaccines and how this 111 varies over time. It is common knowledge that there were progress/positive outcomes and news reportage about the various vaccines, but, there were also more troubling and 113 disconcerting outcomes, especially with the emergence of various variants of the virus. 114 Therefore, the main objective of this research is to explore the exert influence of different 115 news reports about COVID-19 vaccines on people's sentiments, chronologically over one 116 year. The research findings provide valuable insights for policymakers and practitioners by 117 providing an understanding of public sentiment towards COVID-19 vaccines, the impact 118 of news and events on public sentiment, the extent of misinformation about vaccines, and 119 how to develop effective communication strategies. The findings can help policymakers 120 and practitioners to develop more effective strategies to increase vaccine uptake, counter 121 misinformation and protect public health. 122

2. Literature Review

The COVID-19 pandemic has led to a lot of research across the world. Sentiment, 124 Contents, and Retweets: A Study of Two Vaccine-Related Twitter Datasets, published 125 by Blankenship *et al.* [18] in 2018, revealed that leveraging key thought leaders on social 126 media to support health education on vaccination in their tweets might help reach a bigger 127 audience online. The online research focused on how users interacted with the tweets. The 128 research by Deiner et al. [19], found that the public's interest and attitude toward vaccine 129 hesitancy may be seen in social media conversations. The opined that users are more likely to interact with negative news about vaccines. The reason for vaccine scepticism as 131 suggested in [20] might be due to how the government or organization that produces the vaccines communicate the vaccine's benefits. Using Semantic network analysis, the study 133 aimed at raising people's vaccine trust amongst a growing vaccine hesitant population, 134 employing significant concepts to better instruct the target audience. The study suggested 135 that the government should also address concerns that might contribute to vaccine distrust. 136 However, despite the fact that the above studies were effective in establishing how public 137 health communication and social media aid people's scepticism, they failed to address the 138 sways in people's opinions. 139

"How do Twitter users react to TV broadcasts dedicated to vaccines in Italy?" The study revealed that social media discussions in response to public news are priceless and might be used to inform vaccine promotion communication strategies. The study advised that the implementation of a mechanism for monitoring the public's mood toward vaccines on social media should be considered by public health organizations.

Using a logistic regression model on sample data of over one million tweets, Chen and 148 Dredze [22] analysed how images on vaccine-related tweets correlate with the likelihood of 149 it being retweeted and, found that posts with images are twice more likely to be retweeted 150 and shared than nonimage posts. And then recommend that the use of images should be 151 factored into the communication strategies of vaccine administration. While this study 152 was carried out before the COVID-19 outbreak, it highlights the impact of communication 153 strategies on public sentiment with regard to vaccines in general. Another study by Nezhad 154 and Deihimi [23] looked at Persian tweets to understand the Iranian people's view towards 15 COVID-19 homegrown and foreign-made vaccines. Using a sample of over 800,000 tweets 156 posted between April and September 2021, they applied a deep-learning sentiment analysis model based on CNN-LSTM architecture. They found a subtle difference in the Persians' 158 views towards homegrown and imported vaccines; foreign vaccines had higher positive perceptions. While this study clearly underscores the need for positive promotion of 160 vaccines on social media, it did not account for the cultural factors that may impact the 161 overall perception and conversations of Persians around the topic of vaccine administration 162 in Iran. 163

Another COVID-19 sentiment analysis was carried out by Yan et al. [24] on a different 164 social media platform - Reddit, across different cities in Canada, and found three main 165 discussion categories based on topics around Vaccines, vaccine uptakes and vaccine supply. 166 The level of discussion within these topics correlated positively with vaccine acceptance 167 in Canada. This study, just like in [22,23], shows that Social media can be used to better 168 understand sentiments around COVID-19 vaccines and potentially help improve commu-169 nications about vaccines. This finding is in line with the study by Jang et al. [25], which 170 tracked the attitudes of Twitter users in Canada following the vaccine rollouts. In the study, 171 Jang et al. [25] identified two groups of Twitter users who harnessed negative sentiments 172 to achieve divergent goals; the 'anti-vaxxers', who used negative sentiments to discourage 173 vaccine acceptance, and the 'COVID zero' group, who used negative sentiments to criticise the public health response, while encouraging vaccination. Again, another evidence of the 175 impact of social media communication on COVID-19 vaccine participation. It was also discovered in a study by Monselise et al. [26] that fear was a leading of the top 5 emotions 177 from nearly 8 (eight) million tweets collected in a 60-day window that started from December 16, 2020. This study was done in the United States. It is important to remember that 179 some of those sentiments might have been influenced by election misinformation following 180 November 4 2020, US elections [27]. 181

Furthermore, Nearly a million and half unique tweets from over half a million Twitter 182 users were collected by Lyu, Han and Luli [17] between March 2020 and January 2021 183 to understand the changes in public concerns and how they might impact the goal of 184 achieving herd immunity. They found that trust was the most predominant emotion over 185 the period and reached its peak in November 2020, following the announcement of the 186 90% efficacious Pfizer vaccine [28]. This study provided a snapshot of a time during 187 the COVID-19 pandemic when people were, for the first time, required to do physical 188 and social distancing, isolation and quarantine, which had psychological impacts [29]. 189 This means that desperation, fear and hope could have played a huge part in driving 190 public sentiments at that time. According to the study in [30], people's attitudes and 191 emotions towards COVID-19 evolved over time, from positive (at the beginning of the 192 pandemic) to Negative towards the end of 2021. This is further supported by the study 193

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carried out by Niu et al. [31], who attempted to understand the reason for the rapid acceptance of COVID-19 vaccination by the Japanese population at a time when there was a globally diminished confidence in vaccines (January to September 2021), using Twitter comments. They found that negative sentiments outweighed the positive ones, with no increased vaccine confidence, but the communication strategy adopted by the public health authorities helped to create awareness of the danger of COVID-19, which was key to driving up vaccine uptake. 200

3. Methodology



Figure 1. Model Architecture

As can be seen from Figure 1, this study focuses on the extraction and sentiment 202 analysis of tweets (restricted to tweets in English language tweeted from the UK) relating 203 to COVID-19 vaccines, over a period of time. The periods are informed by BBC News re-204 ports/breaking news regarding the vaccines. Therefore, the methodological process of this 205 research starts be identifying periods during the pandemic where BBC News broadcast sig-206 nificant news stories about any of the 3 vaccines administered in the UK - Pfizer-BioNTech, 207 Moderna, and Oxford-AstraZeneca. This is then followed by the extraction of tweets that 208 mentioned any of these vaccines for the identified periods. 209

3.1. Dataset

Six data periods are considered in this research and they are based on BBC News' reportage of significant developments for the UK's adopted COVID-19 vaccines. Each period spans seven days starting from the day the news was reported. The periods and the news headlines are shown in Table 1. The search terms for each of the vaccines are displayed in Table 2. The table shows the terms relating to the three vaccines focused on in this study. It can be seen that some vaccines have more terms than others. These are combinations of terms used to refer to the vaccines to the best of our knowledge.

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Period	News
25 January - 31 January 2021	Moderna vaccine appears to work against variants
2 April - 8 April 2021	COVID 30 blood clot cases found in AstraZeneca
	recipients in the UK
23 May - 29 May 2021	COVID Pfizer and AstraZeneca jabs work against
	Indian variant
9 July - 15 July 2021	Heart inflammation link to Pfizer and Moderna jabs
26 November - 2 December 2021	New COVID variant (Omicron): Javid says UK
	must act quickly over public health risk
8 December 14 December 2021	COVID Vaccines should work against Omicron
o December - 14 December 2021	variant, WHO says

Table 2. Twitter Search Terms

Vaccine	Search Terms
Moderna	(modernavaccine OR modernajab OR modernaCOVIDvaccine OR mod- ernaCOVID19vaccine OR moderna)
AstraZeneca	(AstraZenecavaccine OR AstraZenecajab OR oxfordvaccine OR ox- fordjab OR oxfordCOVIDvaccine OR oxfordCOVID19vaccine OR AstraZenecaCOVIDvaccine OR AstraZenecaCOVID19vaccine OR AstraZenecaoxfordvaccine OR oxfordAstraZenecavaccine OR As- traZeneca)
Pfizer	(pfizervaccine OR pfizerjab OR biontechvaccine OR pfizerbiontech- vaccine OR biontechpfizervaccine OR pfizerCOVIDvaccine OR pfiz- erCOVID19vaccine OR pfizer)



Figure 2. Number of tweets extracted over time

3.2. Data Preprocessing

While extracting the tweets, we made some restrictions and then went further to preprocess the data as follows: 226

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1.	We restricted the tweets to English tweets tweeted from the UK within the timeframes	227
	as per Table 1, and original tweets (retweets are discarded).	228
2.	Hashtags, usernames, and hyperlinks were removed.	229
3.	Special characters were filtered and white spaces removed.	230
4.	The data was tokenised before passing to the BERT model.	231

3.3. Sentiment Detection

In this study, we employed the Bidirectional Encoder Representations from Trans-233 formers (BERT) model to analyse the sentiments in tweets [32]. The BERT model pretrains 234 deep bidirectional representation from unlabeled texts. It considers the contexts of texts by 235 jointly learning from both the left-right and right-left sequences of the texts. Due to this 236 bidirectional representation of the model, stopword removal, stemming, lemmatisation are 237 not useful when using the model as these will affect the texts construct. For example, if 238 we applied stopword removal to the text - the dinner was not good, this will leave us with 239 the text - *dinner good* and the sentiment would have been *positive*. However, the negation 240 not changed the sentiment of the text to negative. The BERT model was trained on 2,500 241 million words from Wikipedia Corpus, and Book Corpus of 800 million words. The model 242 adopts the transformer architecture and uses the following two strategies during training: 243

- 1. Masked Language Model
- 2. Next Sentence Prediction

Input	[CLS] my dog is cute [SEP]	he likes play ##ing [SEP]
Token Embeddings	E _[CLS] E _{my} E _{imaski} E _{is} E _{cute} E _[SEP]	E _{he} E _{play} E _{rring} E _[SEP]
Sentence Embedding	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
Transformer Positional Embedding	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$

Figure 3. Bert Model Process [33]

The Bert model in Figure 3 presents these strategies. Before word sequences are fed into BERT, a part (15%) of the words is replaced with a MASK token. And based on the 247 contexts of the other words (non-masked) in the sequence, the model attempts to predict the masked words. To predict the next sentence (the connectedness between the first and 249 second sentence), the entire input sequence is passed through the Transformer model. The 250 first sentence is the classification sentence [CLS]; all the sentences are separated by the 251 separation token [SEP]. During training, 50% of the time, the second sentence comes after 252 the first, and a randomly sampled sentence 50% of the time. BERT model then predicts if 253 the second sentence is a correct sequence or a random sentence. 254

As earlier stated, the main objective of this research is to explore the exert influence of 255 different news reports about COVID-19 vaccines on people's sentiments, chronologically 256 over one year. The method employed in this research achieved this objective following sim-257 ilar methods as those stated in section 2 which focused on extracting tweets for sentiment 258 analysis. In our study, the chronological analysis of UK-based COVID-19 vaccine related 259 tweets enable us to uncover how news events about the vaccines affected people's opinions. 260 We adopted the BERT model because several studies have shown that BERT outperforms other models in sentiment analysis tasks. For example, in a study by Devlin et al. [32] 262 BERT was trained on a large corpus of tweets and was able to achieve an accuracy of 85% 263 in sentiment analysis, which outperformed other pre-trained models such as ELMO and 264 GPT-2, as well as traditional machine learning models such as SVM and logistic regression. ²⁶⁵ BERT's pre-training on a large corpus of makes it more robust and less prone to over-fitting. ²⁶⁶

4. Results

In this section, we analyse the sentiments contained in tweets towards the three focused vaccines. The periods of this study are based on BBC News reports about various significant vaccine developments from January 2021 to December 2021. 270

4.1. Period 1: 25 January - 1 February 2021

On the 25th of January 2021, the BBC reported that Moderna COVID-19 vaccine 272 appears to work against more infectious variants of the virus, according to the US pharmaceutical company. Although this news was specifically about Moderna, this vaccine had only 65 tweets compared to the 632 tweets for AstraZeneca and 508 for Pfizer. 275



Figure 4. Period 1 - Sentiment percentage of tweets per vaccine

From Figure 4, we see that the sentiments are generally negative for the three vaccines. 276 However, it is interesting to see that the percentages of negative tweets for AstraZeneca 277 and Pfizer are above 50%. Moderna has more positive tweets than the others. 278

Figure 5 shows words that were used in the negative oriented tweets. Although we can see some positive words such as *efficacy*, *good*, *and better*, there are lots of negative words in these tweets. For Moderna, we see words such as *threatening*, *wrong*, *worse*, *delayed*, *and deficient*. AstraZeneca contains negative words such as *failed*, *threaten*, *issue*, *wrong*, *and dispute*. There are other negative sentiments around the *profitability and nationalsim* where vaccine companies are accused of being *profit* focussed, and Governemnt for *nationalism*. Pfizer possessed some negative terms such as *risk*, *threatening*, *problem and failed*.

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(a) Moderna



(c) Pfizer

Figure 5. Period 1 - Words associated with Negative tweets

Over this same period, some of the positives words associated with the vaccines are as 286 follows (see Figure 6): 287





(c) Pfizer

Figure 6. Period 1 - Words associated with Positive tweets

Moderna: amazing, excellent, working, happy, best. AstraZeneca: great, good, thank, amazing, interesting. **Pfizer**: great, thank, happy, excellent, fantastic.

4.2. Period 2: 2 April - 8 April 2021

On the 2nd of April 2021, the BBC reported 30 blood clot cases found in AstraZeneca 293 recipients in the UK. In this period, the number of tweets for Moderna slightly increased from 65 to 92 while those of AstraZeneca and Pfizer decreased from 632 to 415 and from 295 508 to 230, respectively. This is somewhat surprising as one would expect to see more 296 tweets about AstraZeneca due to the negative news about the blood clot. However, the 297 reason for the declining number of tweets about AstraZeneca could be a result of the 298 fading enthusiasm for the vaccine by the British populace, based on a survey of 5,000 299 people, as reported by a Reuters article [34]. The article noted rising unease about the 300 possible link to rare adverse side effects of the vaccine but also highlighted overall high 301 confidence in vaccines in the UK. The article further reported that the AstraZeneca vaccine 302 was suspended by over a dozen of European countries due to reports of blood clots and 303 low platelets in a small number of people who received the vaccine. 304

Overall, for all the vaccines, the percentage of positive sentiments went up slightly 305 while the percentage of negative sentiments decreased, as shown on Figure 7. However, the 306 percentage of negative sentiments for Moderna went up slightly from 46.2% to 46.7% while 307 it's positive sentiment increased from 32.3% to 34.8%. For AstraZeneca, the percentage 308 of negative sentiments went down from 65.5% to 57.1%; while the percentage of positive 309 sentiments went up to 30.1% from 23.1% - this is a significant increase for Astraneneca. 310 The percentage of negative sentiments for Pfizer went down from 59.4% to 50.0%, while 311 the positive percentage went up to 35.2% from 24.0%. Although the sentiments are still 312 generally negative for this period, there's a significant shift in sentiment from negative to 313 positive. 314



(c) Pfizer

Figure 7. Period 2 - Sentiment percentage of tweets per vaccine

Figure 8 shows the word-cloud for negative words associated with the vaccines. Over this period, some of the negative words associated with the vaccines are as follows - we can see some overlaps in the negative words associated with the tweets:

	318
Moderna: problem, clots, dead, effects, blood.	319
AstraZeneca: risk, severe, effect, clot, blood.	320
Pfizer : effect, death, blood, clot, problem.	321



(a) Moderna

(b) AstraZeneca



(c) Pfizer

Figure 8. Period 2 - Words associated with Negative tweets



(a) Moderna



(b) AstraZeneca



(c) Pfizer

Figure 9. Period 2 - Words associated with Positive tweets

Over this same period, some of the positives words associated with the vaccines are as	322
follows (see Figure 9):	323
Moderna : <i>huge</i> , <i>promise</i> , <i>transparency</i> , <i>funny</i> , <i>thanks</i> .	324
AstraZeneca: good, brilliant, safe, happy, benefits.	325
Pfizer : <i>thank</i> , <i>grateful</i> , <i>lovely</i> , <i>good</i> , <i>safe</i> .	326

4.3. Periods 3 to 6

Table 3 presents a summary of sentiment percentages for periods 3 to 6. Generally,328it could be seen that the overall sentiments for each of the vaccines are negative. It is329

important to note that the sentiment percentages are not synonymous with the number of 330 tweets about the vaccines. The table shows that Moderna received less negative sentiments 331 compared to AstraZeneca and Pfizer. It is also observed that AstraZeneca attracted the 332 most negative tweets. At the positive level, Moderna also attracted more positive tweets 333 that AstraZeneca and Pfizer; while Pfizer received a relatively more positive tweets than 334 AstraZeneca. For all the considered periods, period 3 (23 May - 29 May 2021) with the 335 breaking news - COVID Pfizer and AstraZeneca jabs work against Indian variant, received the 336 least negative tweets and the most positive tweets. This is a significant shift in sentiment 337 from period 2 where 30 blood clots cases were found in AstraZeneca recipients. The "good" 338 news reported in period 3, saw Moderna's percentage negative sentiment dropped from 339 46.7% to 29.4%, AstraZeneca from 57.1% to 45.0%, and Pfizer from 50.0% to 43.7%. On the 340 positive side, Moderna's percentage rose 34.8% to 55.9%, AstraZeneca from 30.1% to 43.0%, 341 and Pfizer from 35.2% to 43.4%. These percentages suggest that the news outbreaks may 342 have swayed people's sentiments towards the vaccines.

Period 4 saw another significant shift in sentiments; this time, to the negative. This period (9 July - 15 July 2021) had the news - *Heart inflammation link to Pfizer and Moderna jabs*. In this period, the percentage of negative sentiments for Moderna rose to 42.0% from 29.4%, AstraZeneca to 57.7% from 45.0%, and Pfizer to 51.8% from 43.7%. The percentages of positive sentiments for the three vaccines also dropped from 55.9% to 40.6%, 43.0% to 26.9%, and 43.4% to 38.2% for Moderna, AstraZeneca, and Pfizer, respectively. There are no significant changes for periods 5 and 6. In general, it could be said that people's sentiments shift in relation to breaking news.

Table 3. Periods 3 - 6: Sentiment percentage of tweets per vaccine

Timeline	Negative (%)			Positive (%)		
	Moderna	AstraZeneca	Pfizer	Moderna	AstraZeneca	Pfizer
23 May - 29 May 2021	29.4%	45.0%	43.7%	55.9%	43.0%	43.4%
9 July - 15 July 2021	42.0%	57.7%	51.8%	40.6%	26.9%	38.2%
26 Nov - 2 Dec 2021	45.3%	57.9%	57.9%	35.3%	36.8%	29.8%
8 Dec - 14 Dec 2021	42.0%	57.6%	52.4%	36.2%	24.2%	29.3

5. Discussion

In summary, this paper contributes to knowledge by exploring the general sentiment about COVID-19 vaccines in the UK; chronologically examining the potential effects of news reports on the public sentiment towards COVID-19 vaccines, by comparing the sentiments toward different COVID-19 vaccines in line with these news reports. 353

In section 2, we compare the contributions of related existing studies and the result 357 results of the proposed study. The study in [35] revealed that public opinion about COVID-358 19 vaccinations shifted dramatically over time and across locations. Sentiment analysis 359 can help public health authorities build locally tailored vaccination education initiatives 360 by providing timely insights into public sentiment concerning the COVID-19 vaccine. 361 However, the study only focused on identifying people's sentiment, not considering how 362 external factors such as news affect those tweets. Rahul et al. [36], examined sentiment 363 analysis and topic modelling of COVID-19 vaccine-related tweets from November 1, 2020, 364 to December 16, 2020. The study reported a more positive sentiments than negative 365 sentiments from the over 500,000 tweets that were analyzed. In [37], the authors conducted a study to see how the COVID-19 news influenced Twitter sentiment in four different 367 countries: the United Kingdom, India, Japan, and South Korea. The United Kingdom was found to have one of the highest percentages of negative sentiment. Surprisingly, 369 the United Kingdom was the most affected of the four countries. This study, however, only focused on analysing news headlines without the corresponding impact of these on 371 people's sentiments, especially towards COVID-19 vaccines. Gesualdo et al. [21] tried to 372 answer the question "How do Twitter users react to TV broadcasts dedicated to vaccines in 373 Italy?" They revealed that social media discussions in response to public news are priceless 374

and might be used to inform vaccine promotion communication strategies. While this study correlates with ours, it did not consider a chronological strategy to their analysis. And while their study focused on Italian population, it is also important to conduct a study focused on the UK.

While our study makes important contributions to knowledge and theory, it can also provide valuable insights for policymakers and practitioners in several ways: 380

- Public sentiment towards vaccines: The findings can provide policymakers and practitioners with an understanding of public sentiment towards COVID-19 vaccines in the UK, which can help them to develop more effective communication strategies to increase vaccine uptake.
- Impact of news and events on public sentiment: The findings can provide policymakers and practitioners with an understanding of how news and events impact public sentiment towards COVID-19 vaccines. This can help them to develop more effective communication strategies to address any misconceptions or concerns that people may have about the vaccine.
 Impact of news and events on public sentiment: The findings can provide policymakers are sameled as a sentiment towards COVID-19 vaccines. This can help them to develop more effective sentiment towards cover a sentiment to address any misconceptions or concerns that people may have about the vaccine.
- Misinformation about vaccines: The findings can provide policymakers and practitioners with an understanding of the extent of misinformation about COVID-19 vaccines on social media, which can help them to develop strategies to counter misinformation and protect public health.
- Targeted interventions: The findings can provide policymakers and practitioners with insights into vaccine hesitancy and help them to develop targeted interventions to increase vaccine uptake.
- Development of Communication strategies: The findings can provide policymakers
 and practitioners with insights on how best to communicate the vaccine message to the
 public, the right timing of communication, and the right channels of communication.

6. Conclusion

COVID-19 vaccines are important for the world to get back on track after the ravaging 401 pandemic. However, it was important to note that many people were sceptical of getting 402 vaccinated. Vaccine hesitancy is a global problem and it is important to understand the 403 reasons for this in order to address it. The paper aimed to understand people's sentiments 404 towards the COVID-19 vaccines in the UK, by examining BBC news reports. In particular, 405 we analysed tweets in relation to three COVID-19 vaccines - Moderna, AstraZeneca, and 406 Pfizer over a twelve-month period using six chronological data periods. We found that 407 overall, the Moderna vaccine is generally perceived more positively than the other two vaccines, while the AstraZeneca vaccine attracted more negative sentiments. These percep-409 tions are largely due to the slightly lower risk of Moderna vaccines compared to the others, 410 and the major blood clot concerns of the AstraZeneca vaccine, respectively. 411

The study has achieved the objectives of the research. It explored the general senti-412 ments of the British people about the COVID-19 vaccines. We see that the general sentiment 413 about the vaccines over the study period is more negative. A high vaccination rate is 414 thought to be well-received and potentially lead to high vaccine confidence notwithstand-415 ing the fears and uncertainties about the pandemic. However, we have seen that this was 416 not the case as various news outbreaks which were sometimes positive and negative at 417 other times swayed people's opinions. The news as reported by the BBC were based on the 418 developments of the pandemic and the vaccines. 419

as seen in the shifts in sentiment percentages. This suggests that the media plays a role in shaping public opinion on vaccines. 427

In order to address vaccine hesitancy, it is therefore important to provide accurate 120 and balanced information about the vaccines in the media. Based on the result of this 430 study, there are several things that can be done to manage people's sentiment toward 431 COVID-19 vaccines (and other vaccines) or vaccine hesitancy. First, it is important to 432 provide accurate and updated information about vaccines. Second, it is important to 433 address the concerns of the people. The findings of this study would mean various 434 things to various stakeholders. For the government and policymakers, the understanding 435 that news publications affect people's opinions about events, in this case, COVID-19 436 vaccines, could lead to the implementation of policies that would enable them to monitor 437 the information being conveyed about the events and how people are reacting to it. It would 438 also help them to track vaccine hesitancy and the themes that influence the hesitation, and 430 to implement or adopt better strategies in addressing people's hesitancy. An effective and authoritative information dissemination process would need to be adopted in promoting 441 and strengthening vaccine confidence. Advancing vaccine confidence could be through sustainable engagements with the general public, and partnerships with local authorities, 443 community leaders, influencers, and cultural organisations to promote COVID-19 vaccines. The government would also need to stimulate more responsive, equitable, and accessible 445 programs that reduce barriers to vaccine confidence and uptake. Healthcare practitioners 446 are the most trusted influencers and advisors of vaccination decisions. They can influence 447 people's opinions and sentiments about vaccines, however, they are faced with erratic changes in people's opinions and sentiments mostly due to news reportage and concerns 449 about the vaccines. An understanding of how news outbreaks sway people's opinions and 450 vaccine hesitancy would enable them to implement strategies to manage and support the 451 changing public. 452

It is recommended that future studies be conducted to investigate the impact of the news on other aspects of people's lives, such as their emotions or economic decisions. In addition, future studies could also explore the impact of fake news on people's sentiments and behaviour. This study has limitations as it only focuses on tweets from the UK. In addition, the study only uses a small sample of tweets as the data is collected over a short period of time. Future studies could focus on covering more timelines and include tweets from other countries to understand how social media users perceive COVID-19 vaccines in different countries.

BBC has been used as a source of information in the research because of its wide reach and diffusive characteristics. The BBC has a large audience, and its content is widely shared and distributed through various platforms. Due to its reach and reputation, the BBC's communicative power can be considered relevant. However, we acknowledge that there are other sources of influence on the sentiment of tweets including social media platforms, newspapers, radio, and television. Additionally, individuals are also influenced by their personal experiences, friends, family, and other people in their social network. Therefore, it would be more accurate to state that among the many sources of information that can reach individuals, the BBC has a relevant weight.

Authors' contributions

Conceptualization by Richard Fuzi and Ebuka Ibeke; Methodology by Olasoji Amujo; Software by Ebuka Ibeke and Olasoji Amujo; formal analysis by Ugo Ogara and Celestine Iwendi; Investigation by Ebuka Ibeke and Celestine Iwendi; Resources and data collection by Olasoji Amujo, Richard Fuzi and Ebuka Ibeke; Writing by: Richard Fuzi, Olasoji Amujo and Ugo Ogara; Validation by: Ugo Ogara and Celestine Iwendi.

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References

1.	Reported Cases and Deaths by Country or Territory.	https://www.worldometers.info/	4
	coronavirus/#countries. accessed:28/10/2021.		4
2.	COVID-19 vaccination programme. https://assets.publishing.service.gov.uk/government/		4

- uploads/system/uploads/attachment_data/file/1009174/COVID-19_vaccination_programme_ guidance_for_healthcare_workers_6_August_2021_v3.10.pdf. accessed:28/10/2021.
- 3. Ugochukwu-Ibe, I.M.; Ibeke, E. E-learning and COVID-19: the Nigerian experience: challenges of teaching technical courses in tertiary institutions. CEUR Workshop Proceedings, 2021.
- 4. Krishnan, R.; Nair, S.; Saamuel, B.S.; Justin, S.; Iwendi, C.; Biamba, C.; Ibeke, E. Smart Analysis of Learners Performance Using Learning Analytics for Improving Academic Progression: A Case Study Model. Sustainability 2022, 14, 3378.
- Iwendi, C.; Mohan, S.; Ibeke, E.; Ahmadian, A.; Ciano, T.; et al. Covid-19 fake news sentiment 5. analysis. *Computers and electrical engineering* **2022**, 101, 107967.
- 6. Ibeke, E.; Lin, C.; Wyner, A.; Barawi, M.H. A unified latent variable model for contrastive opinion mining. Frontiers of Computer Science 2020, 14, 404–416.
- 7. Ibeke, E.; Lin, C.; Wyner, A.; Barawi, M.H. Extracting and understanding contrastive opinion 492 through topic relevant sentences. In Proceedings of the Proceedings of the Eighth International 493 Joint Conference on Natural Language Processing (Volume 2: Short Papers), 2017, pp. 395–400. 494
- 8. Vaccination in United Kingdom. https://coronavirus.data.gov.uk/details/vaccinations. accessed:28/10/2021.
- 9. Coronavirus (COVID-19) booster vaccine. https://www.nhs.uk/conditions/coronavirus-covid-19/coronavirus-vaccination/coronavirus-booster-vaccine/. accessed:28/10/2021.
- Na, T.; Cheng, W.; Li, D.; Lu, W.; Li, H. Insight from NLP analysis: COVID-19 vaccines 10. 499 sentiments on social media. arXiv preprint arXiv:2106.04081 2021.
- Marcec, R.; Likic, R. Using twitter for sentiment analysis towards AstraZeneca/Oxford, 11. 501 Pfizer/BioNTech and Moderna COVID-19 vaccines. Postgraduate Medical Journal 2021. 502
- 12. Niu, Q.; Liu, J.; Nagai-Tanima, M.; Aoyama, T.; Masaya, K.; Shinohara, Y.; Matsumura, N. Public 503 Opinion and Sentiment Before and at the Beginning of COVID-19 Vaccinations in Japan: Twitter 504 Analysis. medRxiv 2021. 505
- Hussain, A.; Tahir, A.; Hussain, Z.; Sheikh, Z.; Gogate, M.; Dashtipour, K.; Ali, A.; Sheikh, 13. 506 A. Artificial intelligence–enabled analysis of public attitudes on facebook and twitter toward 507 covid-19 vaccines in the united kingdom and the united states: Observational study. Journal of 508 medical Internet research 2021, 23, e26627. 509
- 14. Adults' Media Use and Attitudes report 2020/21. https://www.ofcom.org.uk/__data/ 510 assets/pdf_file/0025/217834/adults-media-use-and-attitudes-report-2020-21.pdf. ac-511 cessed:28/10/2021. 512
- Charles-Smith, L.E.; Reynolds, T.L.; Cameron, M.A.; Conway, M.; Lau, E.H.; Olsen, J.M.; Pavlin, 15. 513 J.A.; Shigematsu, M.; Streichert, L.C.; Suda, K.J.; et al. Using social media for actionable 514 disease surveillance and outbreak management: a systematic literature review. PloS one 2015, 515 10, e0139701. 516
- Puri, N.; Coomes, E.A.; Haghbayan, H.; Gunaratne, K. Social media and vaccine hesitancy: 16. 517 new updates for the era of COVID-19 and globalized infectious diseases. Human vaccines & 518 immunotherapeutics 2020, 16, 2586-2593. 519
- 17. Lyu, J.C.; Le Han, E.; Luli, G.K. COVID-19 vaccine-related discussion on Twitter: topic modeling 520 and sentiment analysis. Journal of medical Internet research 2021, 23, e24435. 521
- Blankenship, E.B.; Goff, M.E.; Yin, J.; Tse, Z.T.H.; Fu, K.W.; Liang, H.; Saroha, N.; Fung, 18. 522 I.C.H. Sentiment, contents, and retweets: A study of two vaccine-related twitter datasets. The 523 Permanente Journal 2018, 22.
- 19. Deiner, M.S.; Fathy, C.; Kim, J.; Niemeyer, K.; Ramirez, D.; Ackley, S.F.; Liu, F.; Lietman, T.M.; Porco, T.C. Facebook and Twitter vaccine sentiment in response to measles outbreaks. Health 526 informatics journal 2019, 25, 1116–1132. 527
- 20. Kang, G.J.; Ewing-Nelson, S.R.; Mackey, L.; Schlitt, J.T.; Marathe, A.; Abbas, K.M.; Swarup, 528 S. Semantic network analysis of vaccine sentiment in online social media. Vaccine 2017, 529 35, 3621-3638. 530
- 21. Gesualdo, F.; D'Ambrosio, A.; Agricola, E.; Russo, L.; Campagna, I.; Ferretti, B.; Pandolfi, E.; 531 Cristoforetti, M.; Tozzi, A.E.; Rizzo, C. How do Twitter users react to TV broadcasts dedicated 532 to vaccines in Italy? European Journal of Public Health 2020, 30, 481-486. 533

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- 22. Chen, T.; Dredze, M.; et al. Vaccine images on Twitter: analysis of what images are shared. Journal of medical Internet research 2018, 20, e8221.
- Nezhad, Z.B.; Deihimi, M.A. Twitter sentiment analysis from Iran about COVID 19 vaccine. 530 Diabetes & Metabolic Syndrome: Clinical Research & Reviews 2022, 16, 102367. 537
- Yan, C.; Law, M.; Nguyen, S.; Cheung, J.; Kong, J. Comparing public sentiment toward COVID-19 vaccines across canadian cities: analysis of comments on reddit. *Journal of medical Internet research* 2021, p. e32685.
- Jang, H.; Rempel, E.; Roe, I.; Adu, P.; Carenini, G.; Janjua, N.Z.; et al. Tracking Public Attitudes Toward COVID-19 Vaccination on Tweets in Canada: Using Aspect-Based Sentiment Analysis. *Journal of Medical Internet Research* 2022, 24, e35016.
- Monselise, M.; Chang, C.H.; Ferreira, G.; Yang, R.; Yang, C.C.; et al. Topics and sentiments of public concerns regarding COVID-19 vaccines: social media trend analysis. *Journal of Medical Internet Research* 2021, 23, e30765.
- Bolsen, T.; Palm, R. Politicization and COVID-19 vaccine resistance in the US. Progress in molecular biology and translational science 2022, 188, 81.
- Boseley, S.; Oltermann, P. Hopes rise for end of pandemic as Pfizer says vaccine has 90% efficacy. *The Guardian* 2020, 9. https://www.reuters.com/world/uk/uk-survey-finds-rising-uneaseabout-astrazeneca-covid-19-vaccine-2021-04-27/. Accessed:21/10/2022.
- Bozdağ, F. The psychological effects of staying home due to the COVID-19 pandemic. *The Journal of General Psychology* 2021, 148, 226–248.
- Alhuzali, H.; Zhang, T.; Ananiadou, S.; et al. Emotions and Topics Expressed on Twitter During the COVID-19 Pandemic in the United Kingdom: Comparative Geolocation and Text Mining Analysis. *Journal of medical Internet research* 2022, 24, e40323.
- Niu, Q.; Liu, J.; Kato, M.; Nagai-Tanima, M.; Aoyama, T.; et al. The Effect of Fear of Infection and Sufficient Vaccine Reservation Information on Rapid COVID-19 Vaccination in Japan: Evidence From a Retrospective Twitter Analysis. *Journal of Medical Internet Research* 2022, 24, e37466.
- 32. Devlin, J.; Chang, M.W.; Lee, K.; Toutanova, K. Bert: Pre-training of deep bidirectional transformers for language understanding. *arXiv preprint arXiv:1810.04805* **2018**.
- Horev, R. BERT Explained: State of the art language model for NLP. *Towards Data Science* 2018, 562
 10.
- 34. UK survey finds rising unease about AstraZeneca COVID-19 vaccine. https://www.reuters.
 564 com/world/uk/uk-survey-finds-rising-unease-about-astrazeneca-covid-19-vaccine-2021-0
 4-27/. accessed:21/10/2022.
- 35. Liu, S.; Liu, J. Public attitudes toward COVID-19 vaccines on English-language Twitter: A sentiment analysis. *Vaccine* **2021**, *39*, 5499–5505.
- Rahul, K.; Jindal, B.R.; Singh, K.; Meel, P. Analysing Public Sentiments Regarding COVID-19 Vaccine on Twitter. In Proceedings of the 2021 7th International Conference on Advanced Computing and Communication Systems (ICACCS). IEEE, 2021, Vol. 1, pp. 488–493.
- Ghasiya, P.; Okamura, K. Investigating covid-19 news across four nations: A topic modeling and sentiment analysis approach. *Ieee Access* 2021, 9, 36645–36656.