

1 **Positive attitudes towards COVID-19 vaccines: A cross-country analysis.**

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31 **Abstract**

32 COVID-19 severely impacted world health and, as a consequence of the measures implemented to stop
33 the spread of the virus, also irreversibly damaged the world economy. Research shows that receiving
34 the COVID-19 vaccine is the most successful measure to combat the virus and could also address its
35 indirect consequences. However, vaccine hesitancy is growing worldwide and the WHO names this
36 hesitancy as one of the top ten threats to global health. This study investigates the trend in positive
37 attitudes towards vaccines across ten countries, as steps must be taken to improve the trend over time.
38 Furthermore, we investigate those variables related to having a positive attitude, as these factors could
39 potentially increase the uptake of vaccines. We derive our corpus data from vaccine-related tweets,
40 harvested in real-time from Twitter. Using Natural Language Processing, we derive the sentiment and
41 emotions contained in the tweets to construct daily time-series data. We analyse a panel dataset
42 spanning both the Northern and Southern hemispheres from 1 February 2021 to 1 August 2021. To
43 determine the relationship between several variables and the positive sentiment (attitude) towards
44 vaccines, we run various models, including POLS, Panel Fixed Effects and Instrumental Variables
45 estimations. Our results show that more information related to the safety and side-effects of the vaccines
46 are needed to improve positive attitudes towards vaccines. Additionally, government procurement and
47 the vaccine rollout should improve. Accessibility to the vaccine should be a priority, and a collective
48 effort should be made to increase positive messaging about the vaccine, especially on social media. The
49 results of this study are of the utmost importance to policymakers, health workers, and stakeholders
50 who communicate to the public during infectious disease outbreaks. Additionally, the global fight
51 against COVID-19 might be lost if the attitude towards vaccines is not improved.

52

53 **Keywords:** COVID-19; Vaccines; Big Data; Attitudes

54 **JEL classification codes:** C55, I18, I31, J18

55 **1. Introduction**

56 In an attempt to curb the spread of COVID-19, minimise the loss of life and take the pressure off the
57 national health systems, governments worldwide started their vaccine rollout campaigns late in
58 December 2020. However, this rapid rollout of the COVID-19 vaccine has created different emotional

59 responses across the globe. This is problematic since receiving the COVID-19 vaccine is the best
60 possible solution to open up economies and prevent further loss of life. Compounding the problem is
61 the spread of fake news by anti-vaxxers, which has perpetuated the situation by increasing vaccine-
62 hesitancy around the globe (for example, see Sharma [1] and Bonnevie et al. [2]). Spreading fear and
63 anxiety is a significant problem because we know from the existing literature that vaccine efficacy
64 depends not only on the vaccine, but also on the characteristics of the vaccinated (Madison et al. [3],
65 Glaser et al. [4]). Unfortunately, the COVID-19 pandemic has already led to increased depression,
66 loneliness, and stress levels, thereby increasing the efficacy problem (Madison et al. [3]). Adding to
67 this, vaccine hesitancy is growing worldwide and is seen as one of the top ten threats to global health.
68 This makes it easy to see that governments today face a significant challenge.

69 To this end, our primary aim is to conduct a cross-country panel analysis to investigate the trend in
70 positive attitudes towards COVID-19 vaccines over time. This will enable us to determine whether
71 people are becoming more positive towards accepting the vaccine. A secondary aim lies with
72 determining those variables which are significantly related to a positive vaccine attitude and can inform
73 policymakers.

74 Previous studies (Lyu et al. [5], Xue et al. [6], Chopra et al. [7]) analysed the emotions in vaccine-
75 related tweets. However, their primary aim was to better understand the public perceptions, concerns,
76 and emotions related to *COVID-19 vaccine topics and discussions* on social media. They determined
77 the sentiments related to topics and discussions and investigated the strength of discussions and
78 sentiments over time. The main limitations of these studies include that they: i) only analysed English
79 tweets, with no attention being paid to specific geographical areas or comparing the sentiment across
80 different countries, ii) did not use sentiment analysis in further analyses, and iii) did not investigate the
81 variables related to positive vaccine attitudes. We overcome these limitations by constructing a daily
82 time series called the Vaccine Positive Attitude Index (VPAI), a real-time measure of people's positive
83 attitudes toward the COVID-19 vaccine across ten countries for the period 1 February 2021 – 1 August
84 2021. The countries include Australia, Belgium, Germany, Great Britain, France, Italy, the Netherlands,
85 New Zealand, South Africa and Spain.

86 We derive the VPAI using Big Data by extracting a live stream of tweets for specific geographical areas
87 which contains a list of vaccine-related keywords. After the data is cleaned, we use natural language
88 processing to derive the sentiment and emotions of the tweets. After calculating the mean levels of
89 positive sentiment per day, we investigate the trend over time in the VPAI and compare it across our
90 panel of ten countries under investigation. Additionally, we determine which variables are related to the
91 VPAI and, therefore, when addressed, could create a more positive attitude and increase the uptake of
92 COVID-19 vaccines. To limit the effect of confounding factors, we introduce various estimation
93 techniques to address likely endogeneity. We use POLS as a base model and extend the analyses to
94 include Panel Fixed Effects and Instrumental Variables regressions.

95 Our results indicate that the VPAI trends downward over time for the whole sample. If we consider the
96 Northern and Southern hemisphere subsamples, we find the same results. Considering the trends per
97 country, we find in all countries a downward trend except Belgium and the Netherlands, which shows
98 a slightly positive trend. Therefore, it is clear that interventions are needed to change the attitude toward
99 vaccines and increase the uptake. From our modelling, we find those variables that can improve the
100 positive attitude towards the COVID-19 vaccines are information-related to the safety and side-effects
101 of the vaccines, improving trust in vaccines, reviewing regulations implemented to limit the spread of
102 the vaccines as it seems that people weigh-up the benefits of being vaccinated against lockdown
103 regulations. Additionally, increased trust in governments to procure and effectively roll out vaccinations
104 should be a priority. Furthermore, social media platforms, such as Twitter, should launch targeted
105 campaigns focusing on educating people about the safety of vaccines, providing progress on the rollout
106 and encouraging all ages to get vaccinated. We are confident that if the factors found significant in the
107 econometric models (confidence levels of 95 per cent or more) are addressed, the positive attitudes
108 towards vaccines will improve. Policy interventions in line with these recommendations will contribute
109 to the universal plan to restore global health and the world economy.

110 The rest of the paper is structured as follows. The next section contains a brief background of the
111 countries used in our analyses and studies on COVID-19 vaccine hesitancy. Section 3 describes the data

112 and the selected variables, and outlines the methodology used. The results follow in section 4, while the
113 paper concludes in section 5.

114

115 **2. Background and literature review**

116 **2.1 Country background**

117 This study focuses on three Southern hemisphere countries; South Africa, New Zealand and Australia
118 and seven Northern hemisphere countries; Belgium, Germany, Great Britain, France, Italy, the
119 Netherlands, and Spain. Primarily the choice of countries is determined by data availability. However,
120 in future studies, the dataset can be extended to include more countries. The current selection of countries
121 from both hemispheres provides unique insights into people's attitudes to the COVID-19 vaccine. Table
122 1 provides a summary of key facts for each of the countries used in the current study.

123

124

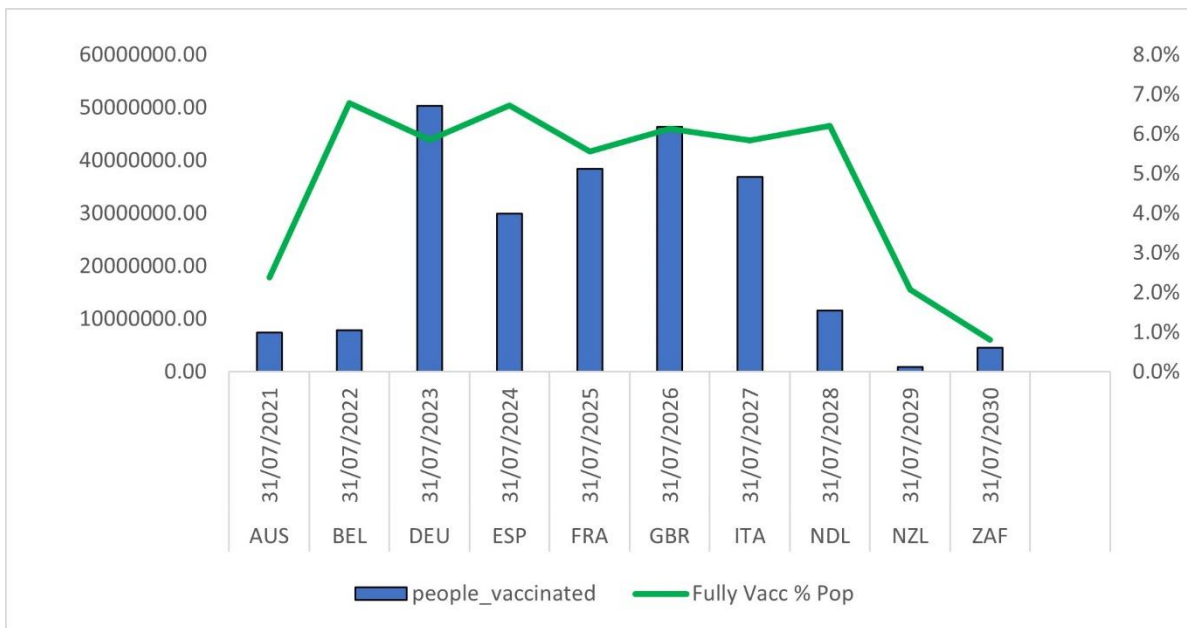
Table 1: Key summary facts of countries in this study

Country	Total population	Average happiness levels^ (2020)	Oxford Stringency Index (Average for the period)	First confirmed COVID-19 case (2020)	Date of first lockdown (2020)	Total confirmed COVID-19 cases (28 August 2021)	Total confirmed COVID-19 deaths (28 August 2021)	Date of vaccine rollout	Percentage of the population fully vaccinated (22 August 2021)
Australia	25.5 million	7.09	58.64	25 January	17 March*	51,256	999	22 February 2021	23.82%
Belgium	11.6 million	6.98	58.30	4 February	13 March	1.18 million	25,360	28 December 2020	67.75%
France	66.99 million	6.66	63.15	24 January	17 March	6.81 million	114,506	27 December 2020	55.57%
Germany	83.02 million	7.08	72.71	27 January	22 March	3.93 million	92,136	27 December 2020	58.56%
Great Britain	66.65 million	7.17	66.70	31 January	23 March	6.73 million	132,699	8 December 2020	61.30%
Italy	60.36 million	6.39	74.90	30 January	9 March	4.52 million	129,002	27 December 2020	58.42%
Netherlands	17.28 million	7.73	64.88	27 February	15 March [†]	1.97 million	18,339	6 January 2021	62.06%
New Zealand	5.5 million	7.14	26.80	28 February	26 March	3,465	26	19 February 2021	20.74%
South Africa	57.7 million	6.32	51.90	6 March	27 March	2.76 million	81,461	17 February 2021	8.05%
Spain	46.94 million	6.40	64.33	31 January	14 March	4.83 million	84,000	27 December 2020	67.27%

* Australia never officially went into a full lockdown such as that seen in the other countries. We used the day when the closure of international borders was announced as a proxy for "lockdown."

[†]The Netherlands started a so-called 'intelligent lockdown' on this date. ^ The happiness scores cited here reflect the average for the period in 2020 before the first COVID-19 case was announced. Sources: Hale et al. [8], Greyling et al. [9], Google [10] [11], Roser et al. [12].

1 From figure 1, it can be seen that the country performing the worst in terms of the total number of
 2 people fully vaccinated is New Zealand (1 million people). However, if we consider the vaccinated as
 3 a percentage of the total population, South Africa performs the worst with 8.05 per cent as of 31 July
 4 2021. Of interest is the Northern-Southern hemisphere split. All three the Southern hemisphere
 5 countries (Australia, New Zealand and South Africa) are outperformed by the countries in the Northern
 6 hemisphere. Of the Northern hemisphere countries, France is the worst performer with 55.57 per cent
 7 fully vaccinated, whereas Belgium is the best performer (67.75 per cent).



8
 9 Source: Hale et al. [8]

10 **Figure 1: COVID-19 number of people vaccinated and the percentage of fully vaccinated people**
 11 **per country (31 July 2021)**

12
 13 **2.2 Literature on COVID-19 vaccine hesitancy**

14 There is an exponential growth of studies in the literature on COVID-19 vaccine hesitancy as
 15 researchers from all disciplines addresses one of the biggest global health threats.

16 Research regarding COVID-19 vaccine hesitancy spans across both *online surveys* (see, for example,
 17 Akarsu et al. [13], Fisher et al. [14], Freeman et al. [15], Ward et al. [16], Seale et al. [17]) and *in-person*
 18 *surveys* (see, for example, Paul et al. [18], Sallam [19]). Primarily these studies found people's hesitancy
 19 and refusal of the COVID-19 vaccine were mostly attributed to i) fear driven by possible side-effects

20 of the vaccine, and ii) the unreliability of what is seen as a new vaccine. Poor compliance with COVID-
21 19 government guidelines was also identified as one of the largest predictors of both COVID-19 vaccine
22 hesitancy and refusal. Additionally, they found willingness to take the COVID-19 vaccine was closely
23 related to one's sense of collective responsibility and campaigning for the 'greater good'. Furthermore,
24 these studies highlighted a need for better and transparent information, the role of anti-vaccination
25 campaigns, and a lack of trust in the government. Interestingly, it was found that low rates of COVID-
26 19 vaccine acceptance were reported in the Middle East, Russia, Africa and several European countries.

27 Our current study uses *Big Data* to construct a Vaccine Positive Attitude Index (VPAI); therefore, the
28 rest of the literature review will focus on those that also use Big Data, with special attention to three
29 studies closest to ours in spirit. We note there is a burgeoning of literature using Big Data in the form
30 of Twitter to analyse vaccine-related topics. Therefore, we cannot possibly discuss all of them.

31 For example, Yousefinaghani et al. [20] used vaccine-related tweets to track frequent hashtags, frequent
32 mentions, main keywords, and main themes with positive and negative sentiments in the tweets.
33 Hussain et al. [21] used Facebook and Twitter to study people's hesitancy, perceptions and sentiment
34 towards the COVID-19 vaccine. Küçükali et al. [22], Nuzhath et al. [23], Bonnevie et al. [2] and
35 Thelwall et al. [24] all identified prominent themes about vaccine hesitancy and refusal on social media
36 during the COVID-19 pandemic. These studies found that the most frequent themes that illicit a
37 negative sentiment are anti-vaccination, poor scientific processes, conspiracy theories, mistrust of
38 scientists and governments, lack of intent to get a COVID-19 vaccine, freedom of choice, and religious
39 beliefs.

40 Sharma et al. [1] and Bonnevie et al. [25] focused on using Twitter to identify suspicious coordinated
41 accounts in the dataset to find misinformation campaigns that drive the conversation against getting the
42 COVID-19 vaccine. Based on an analysis of the collective behaviours and activities of accounts, they
43 found that they correspond to a 'Great Reset' conspiracy theory and ten additional themes such as
44 research and clinical trials and vaccine ingredients.

45 Three studies that come the closest to ours in spirit are Lyu et al. [5], Xue et al. [6] and Chopra et al.
46 [7]. Lyu et al. [5] used 1.5 million English vaccine-related tweets collected between March 2020 and
47 January 2021 and categorised the tweets into 16 topics grouped into five overarching themes. Their
48 results showed that under their first theme called "Opinions and Emotions Around Vaccines and
49 Vaccination", the topic out of all 16 topics that were mostly tweeted was opinions about vaccination. In
50 terms of their sentiment analysis (using the Syuzhet lexicon) they found that, apart from fluctuations
51 throughout the period, the sentiment increased regarding the COVID-19 vaccine. Their emotions
52 analysis (using the NRC lexicon) found trust was the most prevalent emotion, followed by anticipation
53 and fear. They found that before Moderna, one of the first to test their COVID-19 vaccine on humans
54 in April 2020, fear was the most prevalent emotion.

55 Xue et al. [6] analysed 4 million English vaccine-related tweets using a list of 20 hashtags from 7 March
56 to 21 April 2020. Their main aim was to identify popular unigrams (one word) and bigrams (two words),
57 salient topics and themes, and sentiments in the collected tweets. In terms of unigrams, they found
58 "virus", "lockdown", and "quarantine" to be the most popular. Bigrams "COVID-19", "stay home",
59 "corona virus", "social distancing" and "new cases" was the most popular. Furthermore, they identified
60 13 discussion topics from the tweets and categorised them into five different themes. For example,
61 theme 1, "Public health measures to slow the spread of COVID-19", included topics such as face masks,
62 quarantine, test kits, lockdown, safety, vaccine and US shelter-in-place. Their emotions analysis (using
63 the NRC lexicon) showed that anticipation followed by fear, trust, and anger were prevalent across 12
64 of the 13 topics.

65 Chopra et al. [7] collected 1.8 million English vaccine-related tweets from across India, the United
66 States, Great Britain, Brazil, and Australia from June 2020 to April 2021. They aimed to create ten
67 lexical categories, split between two classes, namely emotions (6 categories) and influencing factors (4
68 categories) and study the temporal evolution of these categories across time. The lexical emotions
69 category includes hesitation, sorrow, faith, contentment, anticipation and rage, while their influencing
70 factors are misinformation, vaccine rollout, inequities, and health effects. To measure each category's
71 strength in a given tweet, the authors used the word-count approach. They calculated the strength of the

72 categories monthly and split their period under investigation in two; Before and After the date when
73 each country's government approved the first COVID-19 vaccine. Their results differed across countries
74 with, for example, India experiencing a decrease in the strength of hesitation experienced after vaccine
75 approval, with mentions of health effects contributing the most in tweets with a positive hesitation score.
76 The United States experienced a significant increase in contentment after their vaccine approval. Rage
77 and discussions on misinformation became significantly higher after vaccine approval in India, whereas
78 the opposite was true for the United States.

79 Given the above literature review, no other study has done what we propose to do. We will be the first
80 study to use Big Data to determine the sentiment and emotions related to COVID-19 vaccines through
81 a vaccine positive attitude index. Additionally, no other study has followed the trends in attitudes over
82 time and derived emotion and sentiment time-series data across countries to determine the variables
83 that significantly influence a positive attitude towards the COVID-19 vaccine.

84 **3. Data and methodology**

85 **3.1 Data**

86 In the analyses, we use a cross-country panel dataset with high-frequency daily data (see section 3.2).
87 We analyse the time period from 1 February 2021 to 1 August 2021 (181 days) across ten countries.

88 **3.1.1 Constructing time-series data using sentiment and emotions analysis**

89 To derive our time-series data which capture sentiment and emotions, we construct variables using Big
90 Data by extracting tweets from Twitter. In our analysis we extracted two sets of tweets based on
91 keywords, the one related to COVID-19 vaccines and the other related to government. The tweets
92 containing these words amounted to 1 047 000 tweets. All tweets were extracted according to specific
93 geographical areas (country).

94 The first step in our analysis is to determine the tweets' language (we detected 64 different languages),
95 and all non-English tweets were translated to English. After the translation process, we use natural
96 language processing to extract the sentiment and the underlying emotions of the tweets. To test the

97 robustness of the coding of the sentiment of the translated tweets, we use lexicons in the original
98 language, if available, and repeat the process. We compare the coded sentiment of the translated and
99 original text and find the results strongly correlated.

100 We make use of a suite of lexicons. Each of them differs slightly but with the main aim to determine
101 the sentiment of unstructured text data. The two lexicons mostly used in our analysis are Sentiment 140
102 and NRC (National Research Council of Canada Emotion Lexicon developed by Turney and
103 Mohammad [26]). The other lexicons are used for robustness purposes and are part of the Syuzhet
104 package. The lexicons include Syuzhet, AFINN and Bing. The sentiment is determined by identifying
105 the tweeter's attitude towards an event using variables such as context, tone, etc. and it helps one to
106 form an entire opinion of the text. Depending on the lexicon used, the text (tweet) is coded. For example,
107 if a tweet is positive, it is coded as 0, if neutral 2 and if negative 4.

108 We use the NRC lexicon for the sentiment and emotion analysis to identify the underlying emotions of
109 the tweets. It distinguishes between eight basic emotions: anger, fear, anticipation, trust, surprise,
110 sadness, joy and disgust (the so-called Plutchik [27] wheel of emotions). NRC codes words with different
111 values, ranging from 0 (low) to 8 (high), to express the intensity of an emotion or sentiment.

112 To construct the time-series data, we use the coding of the tweets and derive daily averages. In this
113 manner, we derive a positive sentiment, a negative sentiment and eight emotion time-series (as a
114 robustness test, we derive the sentiment time-series using different lexicons and compare these results
115 using correlation analyses). We perform various additional robustness tests, for example, to determine
116 whether the sampling frequency significantly influences the results. Tweets are sampled, the sentiment
117 is derived, and we construct the time-series data using daily averages. To test the robustness of the
118 frequency we construct the index, we repeat the exercise but construct the time series per hour. We find
119 similar trends in our hourly and daily time series—indicating that the timescale at which sampling takes
120 place does not significantly influence the information obtained. A random sample of tweets per day is
121 extracted to test if the volume directly influences the results. The time series based on these smaller
122 samples (50 per cent and 80 per cent of the originally extracted tweets) is compared to the original time
123 series, and we find that these are highly correlated.

124 **3.2 Selection of variables**

125 **3.2.1 The Outcome Variable: Vaccine Positive Attitude Index (VPAI)**

126 To construct the VPAI index, we follow the method as explained above and extract COVID-19 related
127 tweets using the keywords: *vaccinate, vacc, vaccine, Sputnik V, Sputnik, Sinopharm, Astrazeneca,*
128 *Pfizer (if NEAR) vaccine, Pfizer-BioNTech, Johnson & Johnson, and Moderna.*

129 To ensure that the extracted vaccine-related tweets discuss attitudes related to receiving the COVID-19
130 vaccine, we first constructed word clouds per country. As an example, figure 2 illustrates the word
131 cloud generated for Great Britain. After generating word clouds for all countries, we returned to the
132 original tweets and confirmed the context of the words with high frequencies. We determined that these
133 vaccine-related tweets are indeed related to receiving the vaccine and expressed that "it's good to receive
134 a vaccine" and that people are happy after receiving their second vaccination. For example, tweets that
135 generated the word cloud for Great Britain included:

136 *"Here it is, worth its weight in gold. My consent form for the covid vaccine next week, normality on the*
137 *horizon hope"*

138 *"So excited to hear my mum, an NHS nurse, will be receiving the Pfizer Covid-19 vaccine today - a*
139 *glimmer of hope af"*

140 *"Grandmother the vaccine, as you can see, absolutely delighted ?? (all credit to my younger brother*
141 *for this absolute"*

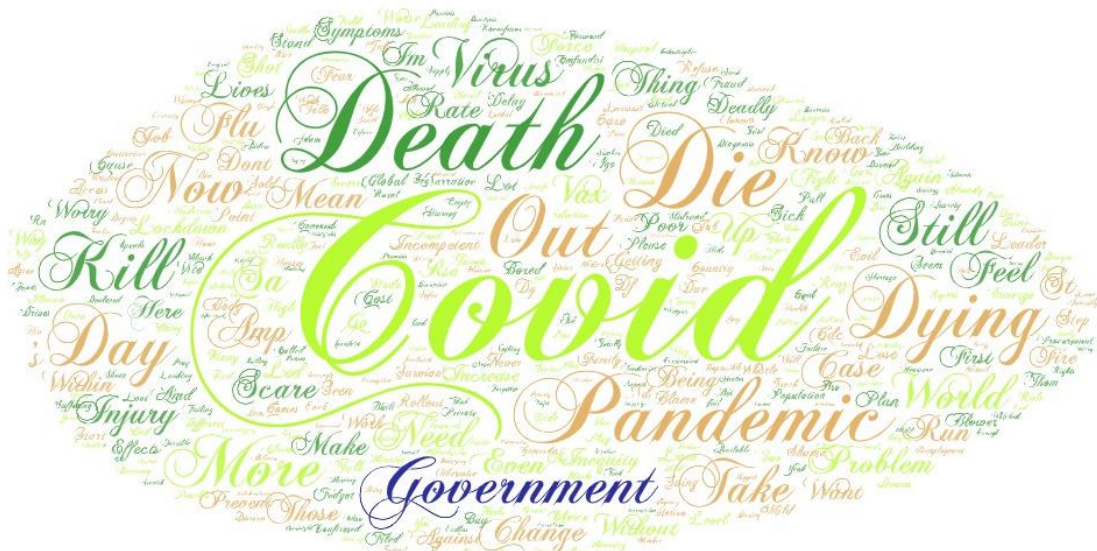
142 Please note the above tweets were taken directly from Twitter and do not represent the views of the
143 authors or their institutions.

164 related to vaccine tweets is not the inverse of the positive sentiment. In our analyses, we find that
 165 negatively coded tweets mostly relates to anger, fear or sadness due to the procurement or rollout of the
 166 vaccines.

167 In terms of theoretical framework, we use a measure that captures relevant predictors of vaccination
 168 behaviour, called the 5C scale. The 5C scale measures the "psychological antecedents of vaccination"
 169 as designed by Betsch et al. [28] and is grounded in established theoretical models of vaccine hesitancy
 170 and acceptance (Thomson et al. [30], MacDonald et al. [31], Larson et al. [32]) and relates these
 171 predictors to psychological models to explain health behaviour (Betsch et al. [29]). Specifically, we
 172 note from the 5C that confidence, constraints, and calculation are important when investigating
 173 vaccination behaviour.

174

175 For the analysis of the negative sentiment and negative emotion vaccine-related tweets, we follow the
 176 same process as described in section 3.2.1. For an example of the issues that cause negative sentiment
 177 and emotions in people, see the word cloud in figure 3 generated from tweets extracted for South Africa.



178

179 Source: Authors own compilation using word cloud software

180 **Figure 3 Word cloud based on negative sentiment for vaccine-related tweets, South Africa.**

181

182 Sample tweets that generated the word cloud for South Africa's negative sentiment include, for example:

183 *"With an incompetent government, a Minister of Health without a medical degree, NDZs dictatorial*
184 *tendencies & a rural population still totally unaware of what a pandemic is added to a vaccine*
185 *shortage, we are doomed"*

186 *We are bored about 1) corruption 2) poor vaccine strategy 3) terrible national government 4)*
187 *incompetent cabinet 5) stealing during a pandemic!"*

188 *This vaccine rollout has been disastrous from the government, has cost lives, now another lockdown*
189 *killing our already hurt economy. Massive change is needed in the running of this country*

190 *The rate with which people are dying every day should get SAHPRA concerned and energised to*
191 *approve more vaccine even on trial basis, this apparent incompetence is really killing and destroying*
192 *families.*

193 Please note that the above tweets were taken directly from Twitter and do not represent the views of the
194 authors or their institutions.

195 After conducting an in-depth analysis of the negative sentiment vaccine-related tweets for all ten
196 countries, we discovered that the negative sentiment was mainly related to anger towards governments'
197 incompetence in procurement, the lack of procuring a sufficient number of vaccines and the execution
198 of the vaccine rollout, fear regarding side-effects, fear of people dying because they cannot get access
199 vaccines and people refusing to be vaccinated. Interestingly, the words prominent in the word cloud,
200 such as 'death', 'die', and 'killing', are related to not receiving the vaccine rather than fearing the side
201 effects.

202

203 In analysing the negative sentiment tweets, we found that tweets expressing dissatisfaction with
204 governments are false negatives. This means that, in reality, people have positive attitudes towards
205 vaccines. Still, they are negative about government incompetence related to issues such as the rollout
206 process, the procurement and accessibility of vaccines etc., hence the negative sentiment of the tweets.
207 To test this hypothesis, we also create a VPAI in which we add the tweets coded as false negatives to
208 the tweets coded as positive. We name the index VPAI2. See the results of the trends in Appendix B.

209 Figures 10 to 12 indicate a predominant upward trend. This suggests that if policymakers address the
210 grievances of people related to the abovementioned government incompetence, they can turn around
211 the downward trend in the VPAI.

212

213 Therefore, the selected covariates included in the regression analyses are:

214 1. Trust in the COVID-19 vaccine: as a proxy for how people perceive the safety of the vaccine.

215 To construct this variable, we follow the method as explained in section 3.3.1. We use the *NRC*
216 *lexicon* to return the emotion score for each COVID-19 vaccine-related tweet for 'trust'. We
217 construct a daily time-series by averaging the measured value of 'trust' per tweet per day
218 (Greyling [9], Betsch et al. [28]). We lag the variable to address likely endogeneity that might
219 spread from confounding factors.

220 2. Anger towards the government: is included in our interaction variable (see point 8). However,

221 to construct the 'anger towards the government' variable, we first extract all tweets that include
222 the following keywords: government, parliament, ministry, minister, senator, MPs, legislator,
223 political, politics, prime minister. We use the same method to construct the time-series as for
224 the 'trust in the COVID-19 vaccine' variable (Greyling et al. [9], Betsch et al. [28]). We use the
225 anger emotion as a proxy for dissatisfaction with the government. We lag the variable.

226 3. Compliance: as a proxy for collective responsibility. We follow Sarracino et al. [33] and define

227 compliance as the degree of association between people's behaviours and COVID-19
228 containment policies to construct the compliance variable. We use information gathered from
229 Google Mobility Reports (the change in duration from the residential category) (Google [10]
230 [11]) and the Stringency Index, which consists of the following nine indicators: school closing,
231 workplace closing, events cancelled, restriction of gatherings, closed public transport, staying
232 at home requirements, restrictions of internal movements, international travel controls, and
233 public information campaigns. The Stringency Index ranges from 0 to 100, with 100 being the
234 most stringent and was sourced from Oxford's COVID-19 Government Response Tracker (Hale
235 [8]). Therefore, we estimate the following equation:

236

$$res_{ct} = \alpha + \beta_{ct} \cdot Country_c \cdot Day_t \cdot Policy_{ct} + \delta_c \cdot Country_c + \lambda_{ms} + \varepsilon_{ct} \quad (1)$$

238

239 where res_{ct} is residential mobility in country c on day t ; $Country$ is a vector of dummies for
 240 each country included in the dataset; Day is a vector of dummies for the days from 1 February
 241 2021 to August 2021. We focus on this period because prior to February 2021, the vaccine
 242 rollout did not occur in all countries under investigation. A vector of dummies is depicted by
 243 λ for each combination of month m and hemisphere s , to account for the different seasons and
 244 evolution of the pandemic among the Northern and Southern hemispheres. The coefficient β_{ct}
 245 is our measure of compliance. It provides the correlation between policy stringency and
 246 mobility by country and day. We are aware that creating a daily compliance measure has the
 247 risk of introducing noise in the correlation. However, to fulfil our aim of determining the daily
 248 evolution of positive attitudes towards vaccines, we do all our analyses daily.

- 249 4. All tweets related to vaccines (Greyling [9]) are a proxy for calculations that refer to individuals'
 250 engagement with COVID-19 content. We assume that individuals highly engaged with
 251 COVID-19 content evaluate the risks of infections and vaccination to make a good decision
 252 (Betsch et al. [28]).
- 253 5. Daily COVID-19 vaccine doses administered per thousand people (Hale et al. [8], Betsch et al.
 254 [28]): a proxy for how well a country's vaccine rollout is being handled. We lag this variable to
 255 address likely endogeneity that might spread from confounding factors. The rollout or lack
 256 thereof also proxies various constraints such as problems with the physical availability of the
 257 COVID-19 vaccine, lack of geographical accessibility, or signalling a less than adequate appeal
 258 for vaccination services uptake. We find that the VPAI and the daily vaccines have an inversely
 259 proportional relationship; therefore, we transformed this variable using a hyperbolic function.
- 260 6. Daily total new cases: a proxy for the evolution of the COVID-19 pandemic across all ten
 261 countries (Hale et al. [8]). In our models, we lag new cases to capture people's expectations of
 262 the trend of the pandemic.

- 263 7. Vaccine policy: we control for the vaccination policy across our ten countries. According to
264 Hale et al. [8], a vaccination policy is classified as follows: 0 - no vaccine available; 1 – vaccine
265 available for one of the following groups: key workers / clinically vulnerable groups / elderly
266 groups; 2 - available for two of the abovementioned groups; 3 - available for all the
267 abovementioned groups; 4 - available for all three groups plus partial additional availability
268 (select broad groups/ages) and 5 – the vaccine is universally available.
- 269 8. To capture anger directed towards the government, we use an interaction variable 'government
270 anger' interacted with 'new daily vaccinations'. This variable captures the anger expressed
271 towards the government given the number of new vaccinations per day. We use the above as a
272 proxy for people's dissatisfaction with the vaccination rollout, which also encapsulates
273 procurement, capacity and corruption issues, and accessibility of the vaccines.

274 Table 2 provides summarised statistics for the variables included in our study.

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Table 2: Descriptive statistics of the variables included in the estimations of attitudes against the COVID-19 vaccine

Variable	Observations	Mean/ Frequency (%)	Std Dev.	Min	Max
VPAI	1,780	0.35	0.12	0.10	0.91
Lagged trust in the COVID-19 vaccine	1,780	0.37	0.09	0.16	0.91
Stringency index	1,780	60.88	17.48	22.22	87.96
Residential mobility	1,780	8.32	24.62	-29.67	50.85
Lagged compliance	1,780	1.78	0.446	1.11	4.10
Lagged anger towards the government # Daily vaccinations	1,780	0.84	0.54	0.00	2.52
Vaccine tweets*	1,780	106.32	108.20	6	690
Lagged new daily vaccinations*	1,780	231776.60	219928.20	0	873515
Lagged new daily cases*	1,780	148	154	0	701
Vaccine policy					
0	605	25.80	-	-	-
1	119	5.88	-	-	-
2	444	21.13	-	-	-
3	574	22.08	-	-	-
4	353	13.98	-	-	-
5	305	11.13	-	-	-

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Source: Authors' calculations. *Note: Statistics on tweets, new daily vaccinations and cases are given in their original format, though in the models, vaccine tweets were logged, and the hyperbolic function of new daily vaccination were derived and lagged; the new daily cases were logged and lagged, and all variables were smoothed.

291 **3.3 Methodology**

292 We use descriptive statistics to analyse the trend in the VPAI over time and compare the results for the
293 Northern and Southern hemispheres and across the ten countries in our sample. Secondly, we use
294 various econometric techniques to derive and test the robustness of the relationships between our
295 selected covariates and the attitudes towards vaccines.

296 The correlation between the VPAI and the covariates over time is likely to be affected by confounding
297 factors, such as the severity of the pandemic, exposure to different types of social media, emotional
298 well-being (depression) of the people, accessibility of the vaccine, the prejudice built into the social-
299 cultural environment and the seasons of the year. Therefore, we resort to various econometric
300 techniques to address biases arising from the confounding effects of these variables.

301 Ideally, we would like to estimate the following equation:

$$302 \quad VPAI_{ct} = \beta_0 + \beta_1 Vac_Trust_{ct-1} + \beta_2 Gov_Anger_{ct-1} + \beta_3 Compliance_{ct-1} + \beta_z X_{ct} + \lambda_m + \mu_c +$$
$$303 \quad \epsilon_{ct} \tag{2}$$

304 where $VPAI_{ct}$ is the vaccine positive attitude index as defined in section 3.2.1 for country c on day t ;
305 Vac_Trust_{ct-1} (see section 3.3.1) is the average level of trust related to the COVID-19 vaccine for
306 country c on day $t - 1$. Gov_Anger_{ct-1} is the average level of anger in government for country c on
307 day $t - 1$; $Compliance_{ct-1}$ is the average level of compliance as defined in section 3.2.2 for country
308 c on day $t - 1$. X_{ct} is a vector of variables, λ_m are monthly effects capturing common effects across
309 countries, such as seasonal effects, trends in the pandemic and holiday seasons, while μ_c are country
310 effects.

311 **3.3.1 Pooled Ordinary Least Squares (POLS)**

312 As a baseline model, we use a POLS estimation. To address the bias that might spread from reverse
313 causality, we lag 'trust of COVID-19 vaccines', 'government anger', 'compliance', 'the daily number of

314 COVID-19 vaccinations' and 'cases'. To address heteroscedasticity, we use robust standard errors in the
315 estimated models.

316 **3.3.2 Fixed Effect (FE) estimation**

317 Having the benefit of a panel dataset allows us to control for additional biases, particularly unobserved
318 confounding factors. Specifically, the FE approach reduces the impact of confounding by time-invariant
319 factors, such as the unobserved and, in this instance, observed characteristics of the countries.

320 To test if the FE model rather than the Random Effects (RE) model is the most efficient estimator in
321 the current study, we use the Hausmann test. We reject the null hypothesis that there is "*no correlation*
322 *between the unique errors and the regressors in the model*", confirming that the FE will give the most
323 robust estimations.

324 The country (individual) FE included in the model addresses the unobserved time-invariant
325 heterogeneity between countries, which considerably reduces the risk of the confounding factors
326 discussed above. Additionally, the FE model also partly addresses bias originating from omitted
327 observed variables (related to country characteristics). However, the FE model cannot address bias for
328 unmeasured time-varying confounding factors or reverse causality. To further address reverse causality,
329 we turn to Instrumental Variable regressions.

330 **3.3.3 Instrumental Variable (IV) regression**

331 In addition to the lagged variables introduced in the POLS and the FE estimations, we also use an IV
332 model to address possible endogeneity and reverse causality. We use the Generalised Method of
333 Moments (GMM) estimation rather than the Two-Stage Least Square (2SLS) estimator, due to the
334 efficiency gains derived from using the optimal weighting matrix. The efficient GMM estimator is
335 robust to the presence of heteroscedasticity of unknown form.

336 We instrument 'lagged trust in the COVID-19 vaccine' and 'lagged compliance', with 'lagged fear of
337 vaccines', 'lagged disgust with the vaccines' and a two-day lag in 'compliance'. We use the Hansen's J
338 statistic to test for over-identifying restrictions. The joint null hypothesis is that the excluded

339 instruments are valid instruments, i.e. uncorrelated with the error term and correctly excluded from the
340 estimated equation. A rejection casts doubt on the validity of the instruments. However, in our specified
341 model, serial correlation is present as the error term in one period is correlated with the errors in previous
342 periods. This causes the estimated variances of the regression coefficients to be biased, leading to
343 unreliable hypothesis testing. Therefore, we consider the IV estimations with the results of the POLS
344 and FE estimations.

345 **4. Results**

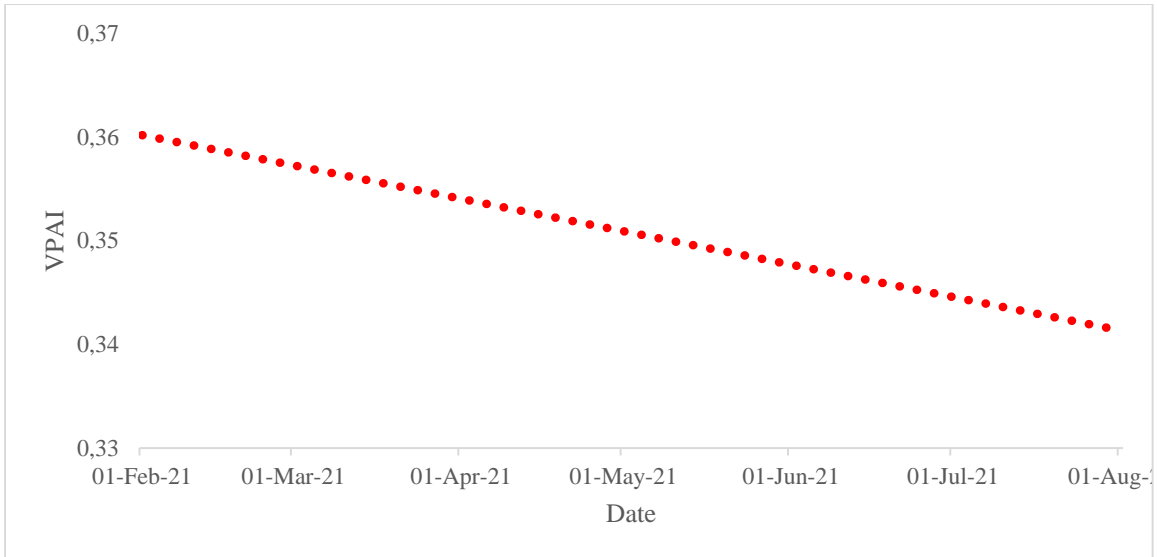
346 **4.1 Results of the trend in the VPAI and emotions**

347 We first focus on our descriptive analysis to explain the trends in the VPAI towards the COVID-19
348 vaccines for the period 1 February 2021 to 1 August 2021. We start by describing the trends in our
349 overall sample, the different hemispheres and then in each country.

350

351 **4.1.1 Trends in the VPAI**

352 When we consider figure 4, we see that the trend in the VPAI towards the COVID-19 vaccines across
353 all countries was downwards; we note an almost 8 per cent decrease over time (see Appendix A for the
354 results using Sentiment 140 rather than NRC). Section 4.1.2 discusses possible explanations for this
355 downward trend.



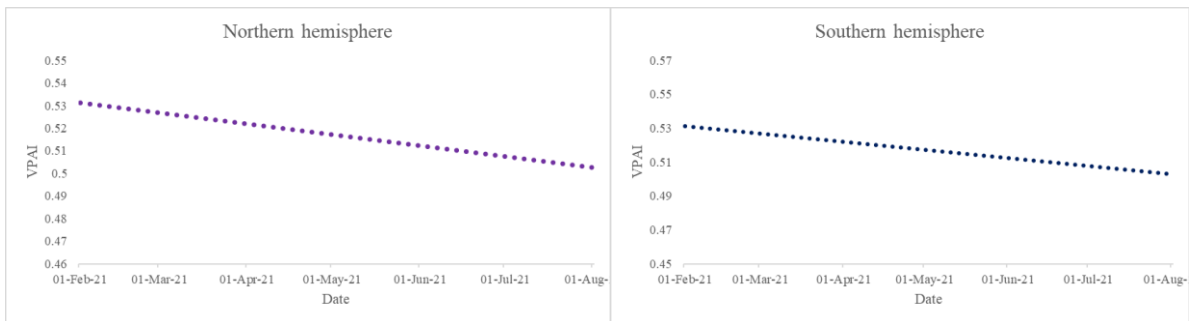
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357 Source: Authors' calculations

358 **Figure 4: Trend in positive attitude from February 2021 to August 2021 for the whole sample**

359

360 Additionally, we note from figure 5 that the downward trend in positive attitude holds across both the
 361 Northern and Southern hemispheres. However, the downward trend seems to be stronger in the Southern
 362 hemisphere than in the Northern Hemisphere.



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364 Source: Authors' calculations

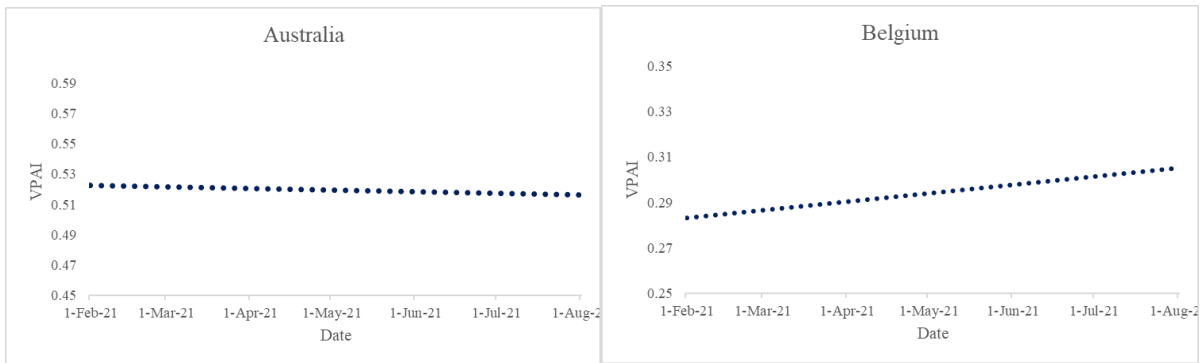
365 **Figure 5: Trend in positive attitude across the Northern and Southern hemispheres from**
 366 **February 2021 to August 2021.**

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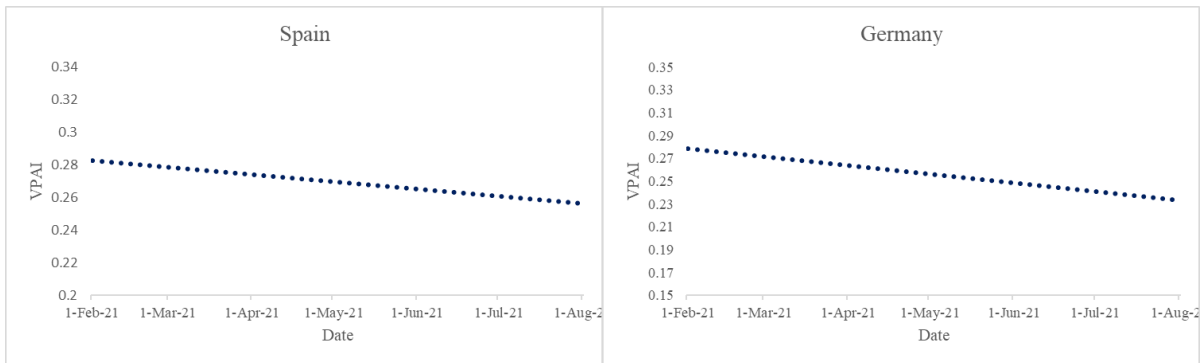
368 4.1.2 Trends in positive attitude per country

369 If we consider the individual countries, figure 6 shows the trend in the VPAI towards the COVID-19
 370 vaccines for each of the ten countries and indicates that the attitude improved in only two countries,
 371 namely Belgium and the Netherlands. For the remaining eight countries, the trend was negative over
 372 time.

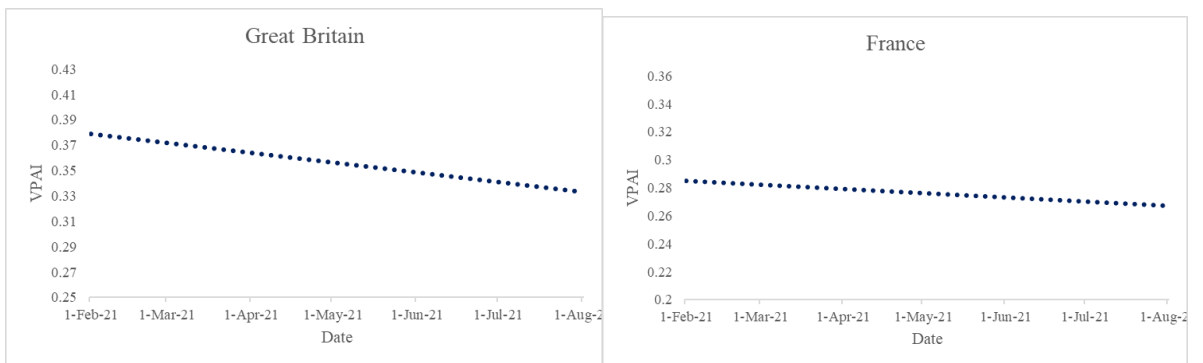
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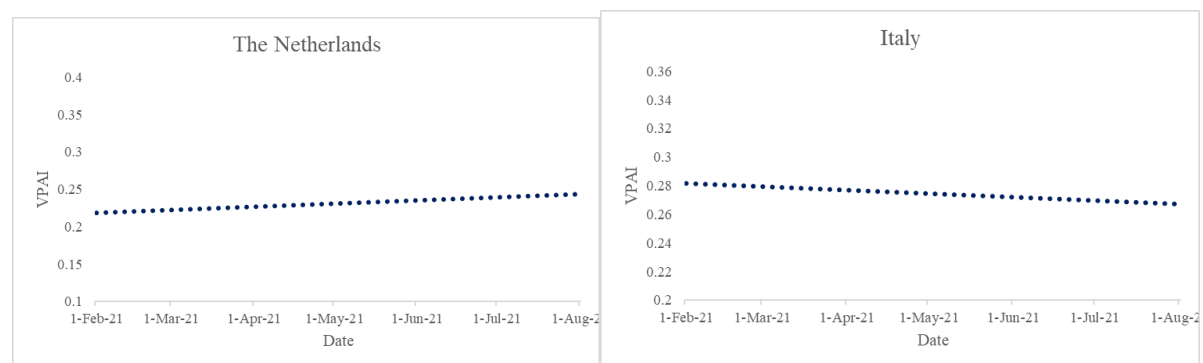
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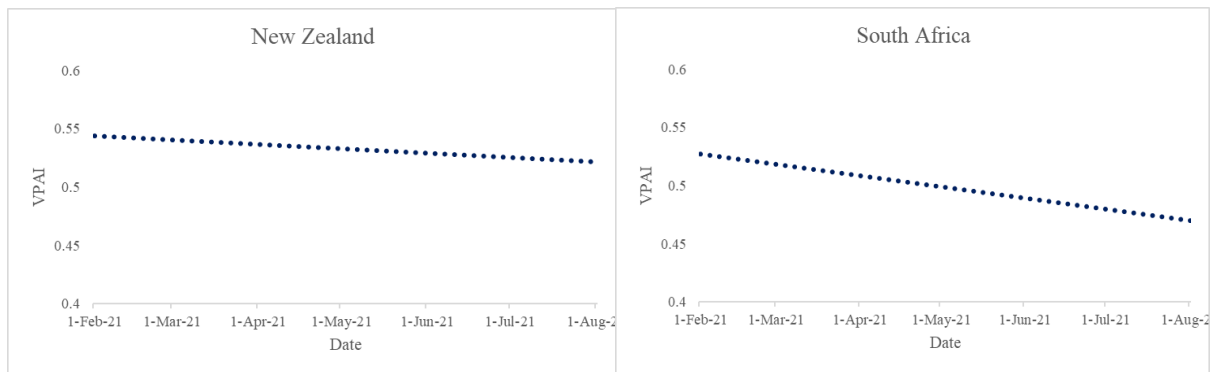
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Source: Authors' calculations

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Figure 6: Trend in the positive attitude for each of the ten countries

381

Upon further investigation, we found that the positive trend in the VPAI in Belgium was likely due to the steps taken to correct government failure that plagued the country in 2020. In 2020, Belgium was the European country with the highest loss of life and hospitalisation rate relative to the size of the population in Europe. Belgium was also hit with capacity issues, struggling to get vaccination centres up and running because of vaccine delivery delays. The Belgian people did not trust information coming from their government after reports of political favouritism in deciding who would get what little vaccine stock was available were leaked (Vanham [34]). The above events led to widespread anger towards the politicians for making COVID-19 a political game. However, many steps were taken to correct the situation, likely turning the attitude towards the COVID-19 vaccine positive. The Belgium government set up a COVID-19 task force responsible for logistics and capacity issues as a first step. However, the high uptake of vaccinations is also the result of lockdown regulation policies. The government decided to relax the lockdown restrictions during spring, depending on the vaccination rates rather than case numbers or hospitalisation rates (Vanham [34]). People wanting to return to 'normal' reacted positively to the policy.

395

The Netherlands was the last European country to start their vaccine rollout on 6 January 2021. Their rollout was hampered by a poor vaccination policy and a conservative strategy that kept more than 40 per cent of its vaccines from being used (Bahceli [35]). However, the Dutch government ended up spending around €6 million on information campaigns in an attempt to increase the vaccine uptake by informing the public about the safety of the various COVID-19 vaccines (Bahceli [35]). On 31 January 2021, a mere 2,430 people had been vaccinated; this number increased significantly, and by 1 August

400

401 2021, 9,288,187 people were fully vaccinated. It is safe to say that the information campaigns have paid
402 off, in the long run, causing a slightly positive trend in the VPAI.

403 Now, let's consider those countries that experienced a decrease in positive attitude over time.
404 Australians have a deep-seated mistrust in their government officials, who seem to implement
405 contradictory regulations across different states. New South Wales (NSW) (Greater Sydney region),
406 home to 32.33 per cent of the total Australian population, has been in lockdown since the middle of
407 June 2021. People living within NSW do not have a great sense of collective responsibility and a
408 significant proportion refuse to comply with government-imposed regulations. In July 2021,
409 approximately 15,000 people protested the lockdowns and demanded that their liberties be restored
410 (Swain [36]). In August 2021 the protests continued, with more than 1,500 police officers disbursing
411 the crowds. Additionally, protestors refuse to wear masks, and the NSW government has collected
412 millions of dollars through fines handed out to people for breaking COVID-19 restrictions. More than
413 \$2.5 million in fines have been collected by the state for non-compliance since the COVID-19 pandemic
414 hit in 2020; however, a staggering 30 per cent - or \$700,000 - has been collected since 19 July
415 (McPherson [37]). Figure 6 shows the downward trend in positive attitudes towards the COVID-19
416 vaccine.

417 Spain shows a downward trend in the VPAI which means that despite vaccination campaigns and a
418 longstanding level of trust that Spaniards have in the public health system, there still is a negative trend.
419 This could possibly be because delivery delays and logistical problem hampered their rollout. In mid-
420 April, when 13 per cent of Great Britain's citizens were fully vaccinated, only about 7 per cent of
421 Spaniards were similarly protected. Future campaigns can rely on the strong family ties of the Spanish
422 and emphasise community benefits of vaccination (AFP [38]).

423 The Germans responded with anger at their government's proposed policy. The policy stated that only
424 vaccinated people would be allowed to enter venues like sports stadiums, movie theatres or restaurants
425 because the residual risk was deemed to be high in such places. In July 2021, Chief of Staff Helge Braun
426 announced that he did not expect another COVID-19 related lockdown in Germany (Schultheis [39]).
427 However, this would mean that if there was a future outbreak, the liberties of the unvaccinated would

428 be taken away with immediate effect. Germany has also seen an increase in people not showing up for
429 their vaccination appointments, with 5 to 10 per cent skipping appointments daily since July 2021. This
430 can be seen as a likely reason for the decreasing trend in the VPAI. Adding fuel to the fire, Germany's
431 anti-lockdown movement, the Querdenken, has been very active in spreading conspiracy theories
432 ranging from the idea that masks are deadly to vaccines being able to alter your DNA (BBC Trending
433 [40]).

434 In Great Britain, the government faces criticism because of their vaccination policy, which has yet to
435 approve the COVID-19 vaccine for 12–15-year-olds. This means sending children back to schools with
436 inadequate mitigations for COVID-19 in place, which could lead to widespread infections and more
437 disruptions to learning. Additionally, trust in the ability of the government to see this pandemic through
438 has decreased since the announcement of their so-called 'Freedom Day' (Donovan [41]). Freedom Day
439 brought with it a lifting of any remaining COVID-19 restrictions and came amidst 47,000 new cases of
440 COVID-19 being reported in the previous 24 hours (Donovan [41]). The decision of Freedom Day
441 brought with it 1,200 scientists worldwide criticising the decision to open up, saying it could pose a
442 threat to the entire world if daily cases increased exponentially and vaccine-resistant mutations of the
443 virus were allowed to develop (Ball [42]). The downward trend in the VPAI is likely (figure 6) a product
444 of all the accumulated issues.

445 In France, introducing a stringent vaccination policy known as 'COVID-19 vaccine passports' has
446 decreased the positive attitude towards the COVID-19 vaccine (The Economist [43]). The news that
447 movie theatres, museums and sports venues have begun asking visitors to provide proof of a COVID-
448 19 vaccination or a negative test has many French nationals angry but willing to take the vaccine simply
449 to return to their once normal way of living.

450 Italy, the second-worst Northern-hemisphere country with regard to people who have not been fully
451 vaccinated, namely 58.42 per cent, has faced an uphill battle since the start of their vaccine rollout on
452 27 December 2020. By April 2021, the Rome government was faced with a lacklustre uptake in
453 COVID-19 vaccines. They decided to take a tough stance, approving emergency legislation to make

454 COVID-19 vaccines mandatory for all healthcare workers, including pharmacy staff (Roberts [44]).
455 This likely contributed to the decreasing trend in the VPAI (see figure 6). Individuals who refused
456 would be transferred to another job or suspended without pay for up to a year. This emergency
457 legislation faced fierce resistance from the country's deeply rooted anti-vaccine movement, which has
458 been fostered in part by populist political forces. These included the 5 Star Movement, which entered
459 government in 2018 promoting vaccine hesitancy. Public trust in the vaccine has also taken a hit after
460 the country temporarily decided to suspend the use of the Oxford/AstraZeneca vaccine after several
461 deaths (Roberts [44]).

462 In New Zealand, the COVID-19 rollout has been described as a 'shambles'. This is due to the
463 government of the day not delivering the promised vaccines because they and the country as a whole
464 became complacent (Vance [45]). At the time of writing this paper, New Zealand found itself in a level-
465 4 lockdown (the most stringent level of lockdown) and had not had a positive COVID-19 case during
466 the previous six months. New Zealand's zero COVID-19 strategies were successful until the first Delta-
467 variant positive person was announced. Soon the government realised that they did not have enough
468 vaccines to vaccinate everyone fully, as previously promised. This government failure has driven an
469 increased mistrust in institutions and anger towards capacity issues in New Zealand. Additionally,
470 people expressed anger towards the government's vaccination policy, making it mandatory for all
471 'frontline' staff to be fully vaccinated or removed from their positions. All of the above can affect the
472 downward trend in the VPAI.

473 South Africa's woes are almost too many to count, ranging from capacity issues, mistrust in the
474 government to anti-vaccination campaigns, all of these creating a decrease in positive attitude towards
475 the COVID-19 vaccines (see figure 6). From as early as December 2020, it was clear that the country
476 had no COVID-19 strategy apart from its dependency on their fragile COVAX arrangement. After
477 receiving their first delivery of the AstraZeneca vaccine on 1 February 2021, it became clear that the
478 government also did not have a clear vaccination policy (van den Heever et al. [46]). The Health
479 Minister created confusion in the public arena when he announced that the AstraZeneca vaccine did not
480 demonstrate efficacy against mild to moderate COVID-19 and placed the rollout of the vaccine on hold.

481 The announcement by the Health Minister caused a decrease in trust in the COVID-19 vaccine and
482 likely contributed to the downward trend in the VPAI. The decision was criticised by local scientists
483 and not supported by the World Health Organisation. This move guaranteed that South Africa would
484 face a winter epidemic wave with most of the 17 million or so high-risk population unvaccinated (van
485 den Heever et al. [46]). During the winter months from June to September 2021, South Africa lost
486 25,660 lives to COVID-19. This probably could have been avoided if the South African government
487 had not been plagued by corruption and mismanagement during its response to the pandemic. By August
488 2021, South Africa saw 'vaccine apathy' or 'vaccine fatigue', with the number of people coming forward
489 to be vaccinated dropping below 200,000 a day, falling short of the set target of 300,000. According to
490 a study conducted by the Human Sciences Research Council and the University of Johannesburg
491 (Cooper et al. [47]), the vaccine-hesitant cite three primary concerns: side effects, effectiveness, and
492 distrust of the vaccine and institutions.

493 To summarise, the downward trend in positive attitudes is partly due to a fear of the side effects, but
494 many other factors also contribute. These include dissatisfaction with governments' rollout plan,
495 procurement, and corruption.

496

497 **4.2 POLS, FE and IV regression results**

498 This section discusses the results of those covariates that are significantly related to the VPAI and,
499 therefore, when addressed, could increase the uptake of COVID-19 vaccines.

500 In table 3, the results of the POLS estimation controlling for a month and a country fixed effect are
501 similar to the results of the FE model and the IV regression. The covariate 'trust in the COVID-19
502 vaccine' is statistically significant and positively related to VPAI across all the estimated models.
503 Therefore, we believe that this result on the correlation between trust and the VPAI is robust. We assume
504 that when trust in the vaccine increases, the fear of negative side-effects decreases and that the positive
505 attitude towards vaccines improves.

506 **Table 3: Results from POLS with FE and IV**

Variable	POLS		FE		IV	
	Coefficient	SE	Coefficient	SE	Coefficient	SE
VPAI						
Lagged trust in the COVID-19 vaccine	0.2938***	(0.0281)	0.2938***	(0.0193)	0.3127***	(0.0999)
Lagged compliance	-0.0176***	(0.0060)	-0.0176***	(0.0037)	-0.0170***	(0.0064)
Lagged anger towards the government # Daily vaccinations	-0.0274***	(0.0073)	-0.0274***	(0.0073)	-0.0273***	(0.0072)
Lagged new daily vaccinations	0.0055***	(0.0013)	0.0055***	(0.0014)	0.0055***	(0.0014)
Lagged new daily cases	-0.0032***	(0.0080)	-0.0032***	(0.0011)	-0.00324***	(0.0080)
Log vacc tweets	-0.0049*	(0.0029)	-0.0049*	(0.0029)	-0.0047*	(0.0029)
Vaccine policy (Reference – Level 0)						
Level 1	0.0598***	-0.0125	0.0598**	-0.0232	0.0596***	(0.0123)
Level 2	0.0451***	-0.0091	0.0451**	-0.0227	0.0454***	(0.0105)
Level 3	0.0476***	-0.0088	0.0476**	-0.0228	0.0480***	(0.0092)
Level 4	0.0536***	-0.0092	0.0536**	-0.0227	0.0542***	(0.0103)
Level 5	0.0473***	-0.0102	0.0473**	-0.0233	0.0477***	(0.0109)
Country FE	Yes		Yes		Yes	
Month FE	Yes		Yes		Yes	
N	1727		1727		1727	
Adjusted R ²	0.867		0.422		0.866	
Hansen J-Statistic of overidentification					p = 0.6544	

507 Source: Authors' calculations
 508 Robust Standard errors in parentheses * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

509

510 Compliance, the act of complying with government-mandated regulations to curb the spread of COVID-
511 19, is statistically significant and negatively related to the VPAI. When people are unwilling to comply
512 with the regulations, their attitude towards vaccines is more positive. Unwillingness to comply likely
513 motivates people to get vaccinated. This finding is important for policymakers because it is necessary
514 to emphasise that the regulations can be less strict once people are vaccinated, returning to a more
515 'normal' way of life, encouraging people to take the vaccine.

516 The variable 'anger towards the government' interacted with the new daily vaccinations per 1000 is
517 statistically significant and negatively related to VPAI. Therefore, when people's dissatisfaction with
518 the government increases, given the incompetence regarding the vaccine rollout, the relationship with
519 the VPAI decreases. Analysing the negative sentiment and emotion tweets that reflect anger, we find
520 that people are angry with governments due to the lack of procurement, procurement of the correct
521 vaccine, the rollout of the vaccination plan and corruption within governments. This anger directed at
522 governments due to a lack of access to vaccines sabotages the positive attitude towards vaccines and
523 hinders the uptake of vaccines.

524 The relationship between VPAI and the number of daily new vaccinations is inversely proportional,
525 significant and positively related. This implies that when the daily number of vaccines administered are
526 very low, the positive attitude is high, but as the number of vaccines administered per day increases,
527 the positive attitude starts to plateau. Also, using the vaccine rollout or the lack thereof, as a proxy for
528 constraints in information campaigns, the physical availability of vaccines or a lack of geographical
529 accessibility, we can see how important it is to overcome any barriers which might impede the intention
530 to be vaccinated (Cylus & Papanicolas [48]).

531 We find that daily new cases, a proxy for the evolution of the COVID-19 pandemic across all ten
532 countries, is statistically significant and negative. If the daily cases are high, the positive attitude
533 towards the vaccine is relatively low, but as the daily cases start increasing, the positive attitude
534 improves.

535 Controlling for the vaccination policy, thus the groups that can access the COVID-19 vaccine, we find
536 that when more groups of people can access the vaccine, for example, all age groups compared to fewer
537 groups, it is positively related to the VPAI. Once again showing that when more people have access to
538 the COVID-19 vaccine, positivity towards the vaccine is enhanced.

539 The number of vaccine-related tweets is statistically significant and negatively related to the VPAI in
540 all the estimated models. This implies that, as the tweets related to vaccines increase, the attitude
541 towards vaccines decreases. This may likely be because many of the tweets contain misinformation or
542 conspiracy theories rather than campaigns and information to encourage being vaccinated - and,
543 therefore, decreases positive attitudes towards vaccines. From this, we can derive that calculation
544 (individuals' engagement in information searching or sharing) plays a significant role in evaluating the
545 risks of infections and receiving vaccinations when making good decisions (Brewer et al. [49]).
546 Depending on the sources of information consulted, high levels of information searching could lead to
547 non-vaccination, due to the abundance of anti-vaccination sources, for instance, on the internet (Kata
548 [50]).

549 In summary, we see that those variables that can improve the positive attitude towards vaccines are
550 related to information about the safety and side-effects of the vaccines (increased trust in vaccines) and
551 a balance between the strictness of regulations and access to vaccines. Additionally, increased trust in
552 the governments' capabilities, honesty of governments and capacity constraints can decrease the
553 dissatisfaction with governments and increase vaccine uptake. Better information about the COVID-19
554 vaccines in general also disseminated via social media can increase positivity towards the COVID-19
555 vaccine. Misinformation about COVID-19 vaccines and social media should be monitored, and
556 campaigns against this misinformation should be launched. Vaccines should also be made accessible to
557 all groups of people.

558 **5. Conclusions**

559 In this study, we constructed a real-time Vaccine Positive Attitude Index (VPAI) derived from Big Data,
560 to illustrate the evolution of people's positive attitudes toward the COVID-19 vaccine across ten

561 countries. Our descriptive analysis showed that the VPAI generally indicates a decline in attitude over
562 the time period investigated. When we consider the different hemispheres, the trend is downwards in
563 the Northern and Southern hemispheres. When we examined the ten individual countries, we saw that
564 only Belgium and the Netherlands experienced a positive trend in the VPAI, whereas the other countries
565 experienced a negative trend.

566 Using POLS, FE and IV regression models, we determined which variables are significantly related to
567 the VPAI and, therefore, could increase the uptake of COVID-19 vaccines if addressed by policy
568 measures. We found that those variables that could improve people's attitudes towards vaccines were
569 information related to the safety and side-effects of the vaccines, increased confidence in governments
570 in conducting the vaccine rollout and handling procurement and capacity issues, cognisance of the
571 compliance versus the vaccine up-take decision, and better information about the COVID-19 vaccines
572 in general, but especially disseminated via social media.

573 These results give policymakers the necessary information on how to increase positive attitudes towards
574 the COVID-19 vaccine. Policymakers should focus on increasing trust in the COVID-19 vaccines. They
575 could more openly disseminate information regarding the vaccine, do it in layman's terms, and
576 acknowledge people's fears, anger, and other negative emotions while emphasising the stringent safety
577 and efficacy standards of the COVID-19 vaccine development process thus fostering individuals' self-
578 efficacy through vaccination. All of this may increase vaccine confidence. Additionally, addressing the
579 lack of accessibility to vaccination clinics could be overcome by following countries such as the United
580 States of America by creating mobile vaccination clinics, to reach people in remote areas.

581 Additionally, policymakers should implement policies to increase people's sense of collective
582 responsibility. This can be done by raising awareness of emotional manipulations by anti-vaccine
583 disinformation efforts and activating positive emotions such as altruism and hope as part of vaccine
584 education endeavours. Another potential strategy is to elicit positive emotions toward helping one's
585 community restore health and well-being, when deciding to vaccinate against what is called the most
586 consequential disease of our time.

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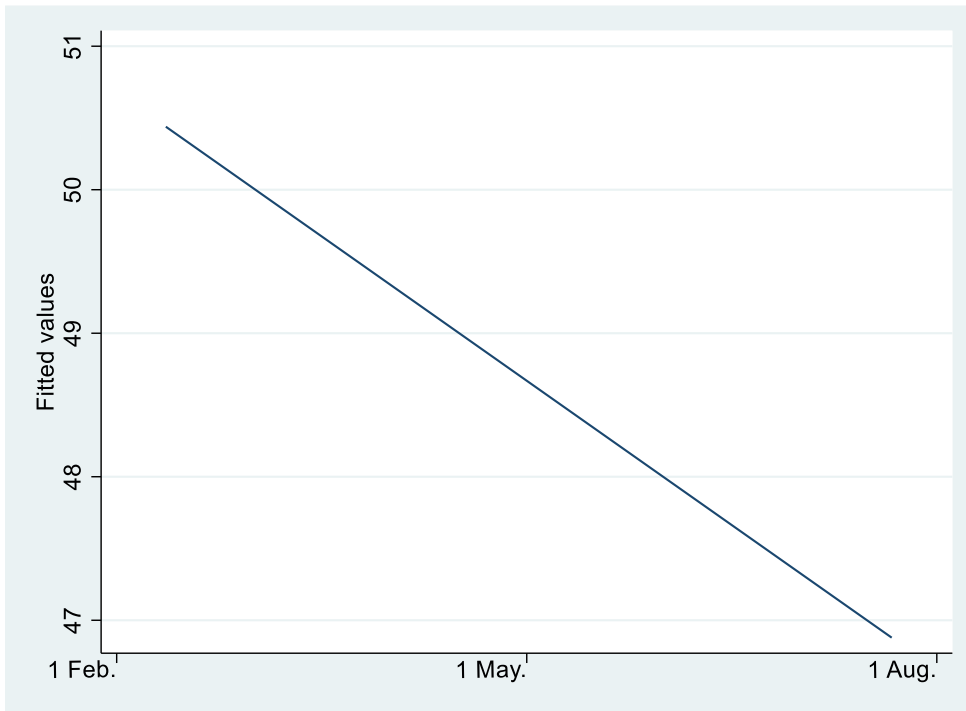
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730

731 **Appendix A: Results on the trends in VPAI using sentiment 140**

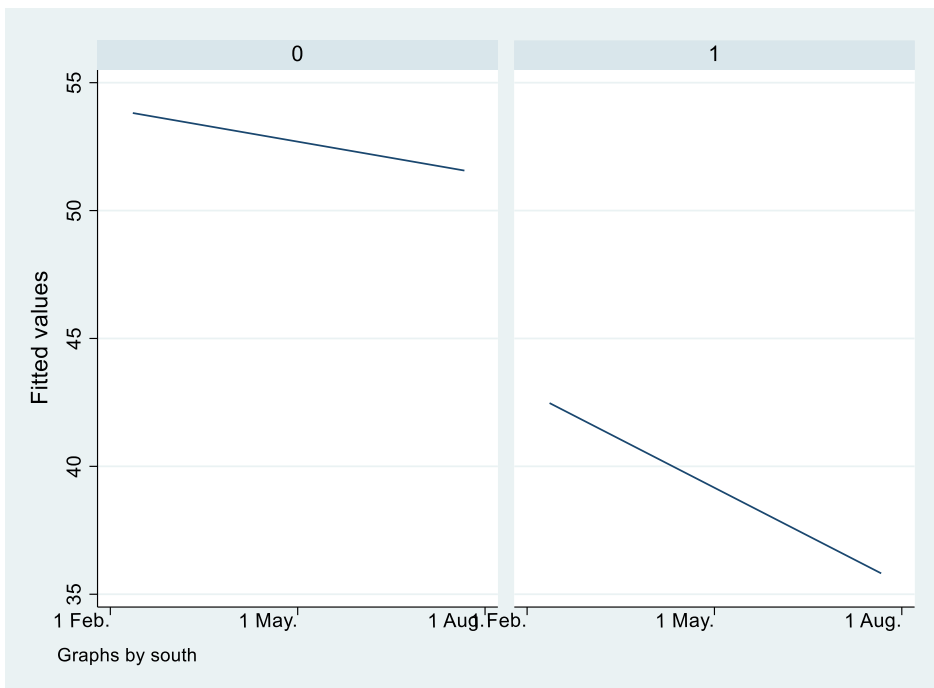


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733 Source: Authors' calculations

734 **Figure 7: Trend in positive attitude from February 2021 to August 2021 for the whole sample,**
735 **using Sentiment 140**

736

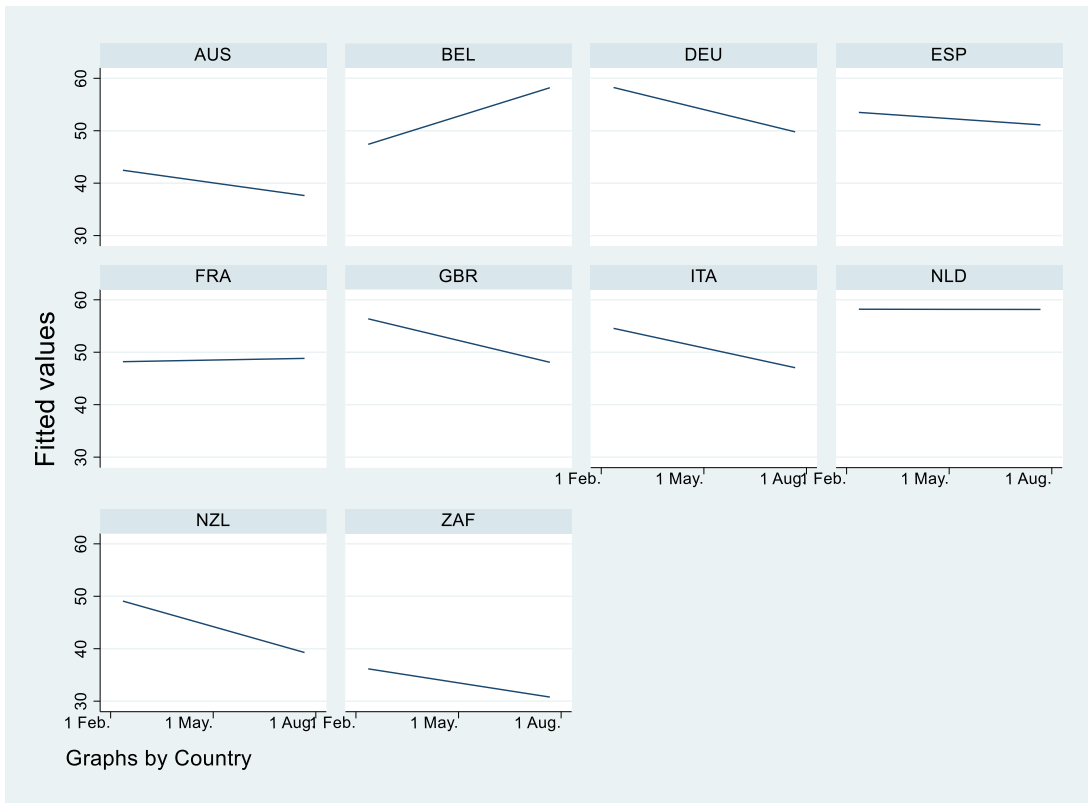


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738 Source: Authors' calculations

739 **Figure 8: Trend in positive attitude from February 2021 to August 2021 per hemisphere, using**
740 **Sentiment 140**

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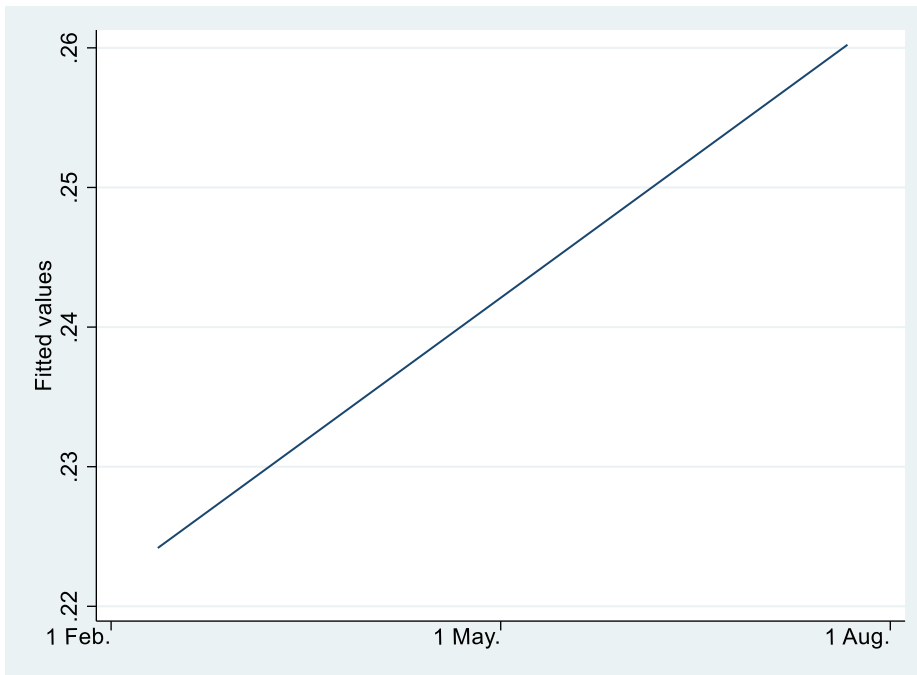
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743 Source: Authors' calculations

744 **Figure 9: Trend in positive attitude from February 2021 to August 2021 per individual country,**
745 **using Sentiment 140**

746 **Appendix B: Results on the trends using VPAI2**

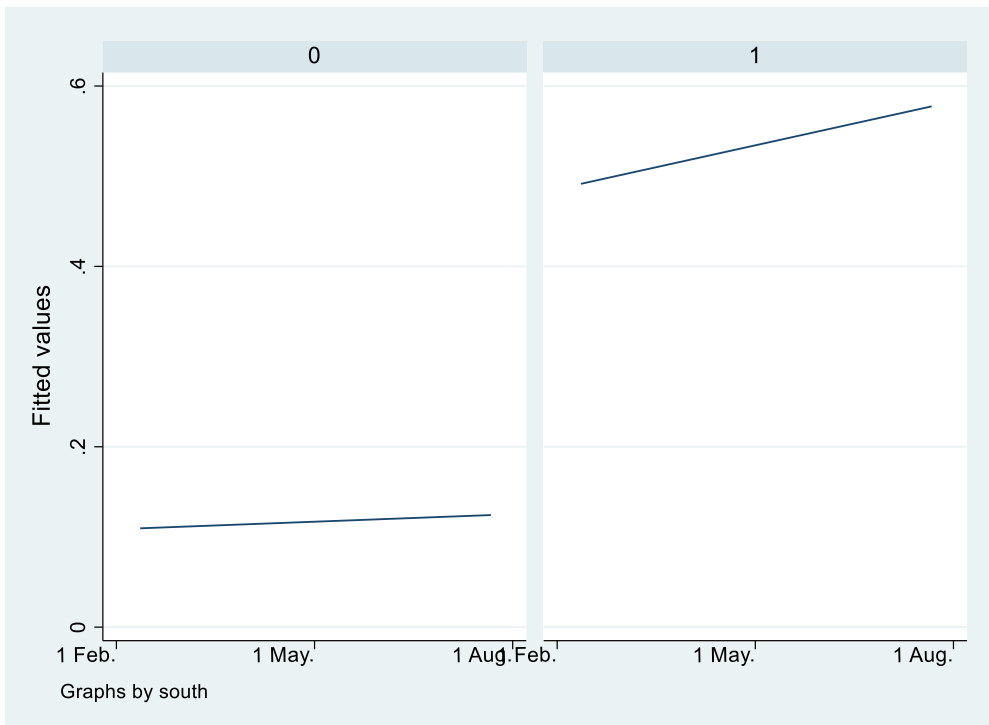
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749 Source: Authors' calculations

750 **Figure 10: Trend in positive attitude including false negatives related to governments (VPAI2)**
751 **for the whole sample from February 2021 to August 2021**

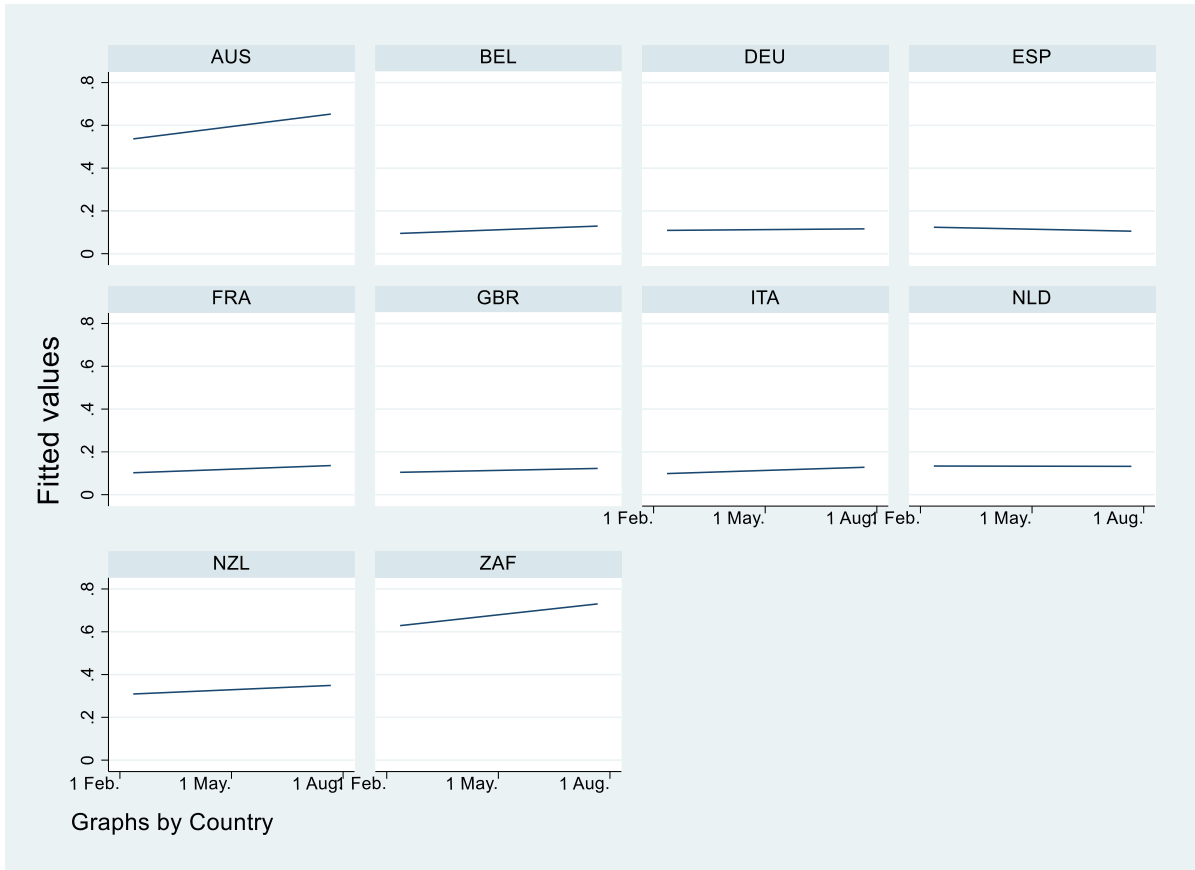


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753 Source: Authors' calculations

754 **Figure 11: Trend in positive attitude including false negatives related to governments (VPAI2)**
755 **for the different hemispheres from February 2021 to August 2021**

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758 Source: Authors' calculations

759 **Figure 12: Trend in positive attitude including false negatives related to governments (VPAI2)**
 760 **for the individual countries from February 2021 to August 2021**