



*Research Article*

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## The Twitter origins and evolution of the COVID-19 “plandemic” conspiracy theory

*Tweets about “plandemic” (e.g., #plandemic) – the notion that the COVID-19 pandemic was planned or fraudulent – helped to spread several distinct conspiracy theories related to COVID-19. But the term’s catchy nature attracted attention from anti-vaccine activist filmmakers who ultimately created Plandemic the 26-minute documentary. Plandemic falsely attacks NIAID Director Dr. Anthony Fauci, among others, and an eventual coronavirus vaccine. The film, which has since been widely discredited, appeared to at least temporarily shift Twitter communications to different topics and organizations, fueling the flow of conspiracy theories and misinformation itself with specific public figures to demonize.*

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### Research questions

- How did the May 2020 release of the documentary *Plandemic* change ongoing Twitter discourse mentioning the term “plandemic”?
- What are the characteristics of “plandemic” tweets before versus after the film?
- Which tweet characteristics are associated with more likes and retweets before versus after the film?

### Essay summary

- On May 4<sup>th</sup>, 2020 the first half of the documentary *Plandemic* was released and rapidly spread via social media by leveraging pre-existing cynicism about COVID-19. Given heightened global attention on preventing the spread of coronavirus and treating COVID-19, misinformation about the origins of the virus and how to stop it warrant immediate attention, so as to protect public health.
- Prior to May 4<sup>th</sup>, “plandemic” (e.g., #plandemic, planDEmic) was a frequently used social media

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term associated with several popular conspiracy theories. By harnessing a pre-existing social media term/hashtag, the film’s producers and promoters had access to captive online networks of conspiracy theory believers through which to initiate the distribution of the documentary.

- We collected 84,884 publicly accessible tweets mentioning “plandemic” between January 24<sup>th</sup> (first mention on Twitter) and May 17<sup>th</sup> (two weeks after the film’s May 4<sup>th</sup> online release). The content and popularity of tweets were analyzed before and after the documentary’s release.
- Twitter discourse mentioning “plandemic” spiked after the film’s release but receded to observed pre-film levels within a 2-week post-film period. Specifically, vaccine-related tweets were relatively marginal and unaltered by the film’s release. In addition, the film increased attention towards certain political, public health, media organizations, and perceived elite public figures. Discourse about freedoms and liberties also increased after the film’s release.
- Future research must explore underlying drivers of anti-science and anti-evidence sentiment by learning about the priorities and health beliefs of people who subscribe to conspiracy theories like the plandemic conspiracy theory and are influenced by and share similar misinformation. Public health advocates and health educators must also participate through preventive interventions and campaigns.

## Implications

The COVID-19 pandemic has resulted in hundreds of thousands of lives lost globally. It has also facilitated the spread of misinformation and conspiracy theories at a scale and pace that is unprecedented (Cuan-Baltazar et al., 2020; Fisher, 2020), leading many to refer to this mis- and disinformation crisis as a “misinfodemic”, defined as the “viral spread of false information” (McGinty & Gyenes, 2020). COVID-19 conspiracy theories have arisen from the fringes of social media conversations to discussions on mainstream media (Ahmed et al., 2020; Uscinski et al., 2020). Understanding the origins, motivations, and evolution of viral misinformation is a key function of social media surveillance, and timely research is needed to combat misinformation’s spread (Ahmed et al., 2020; Y. K. Chang et al., 2020; Chou et al., 2018). In the current study, we analyzed Twitter posts about the plandemic conspiracy theory (i.e., planned epidemic) as a case study for the emergence, evolution, and dynamism of COVID-19 misinformation specifically, and social media misinformation in general. We identify next steps with applications at multiple levels including for researchers, public health practitioners, and social media platforms.

### *The Plandemic conspiracy theory*

On May 4<sup>th</sup>, 2020, the 26-minute film *Plandemic* (“the film”, “the documentary”) was released online and shared via Twitter, YouTube, and other social media platforms (Andrews, 2020; Shepherd, 2020). Prior to May 4<sup>th</sup>, “plandemic” was a frequently used social media term and hashtag associated with several popular conspiracy theories, the general gist of most being that the pandemic was fake (i.e., virus does not exist) or human-made. Driven by expert-like testimony and deliberately “viral [...] shocking and conspiratorial” branding (Rottenberg & Perman, 2020), the documentary leveraged underlying beliefs about COVID-19 and anti-containment sentiments and diverted viewers towards anti-vaccine behaviors and general questioning the pandemic’s impact on our freedoms and liberties (Frenkel et al., 2020). Specifically, the documentary aimed to expose baseless accusations of corruption (Alba, 2020) among key experts in the pandemic response (e.g., Dr. Anthony Fauci) while also suggesting broader collusion among politicians (e.g., Barack Obama) and global elite (e.g., Bill Gates). In other words, the film politicized and demonized public health figures combatting the pandemic. The film’s rapid spread was facilitated by pre-

existing social media networks, such as Facebook groups and viral hashtags (Frenkel et al., 2020). Uniting disparate fringe-belief groups is a common tactic among vaccine opponents (Kata, 2012).

Throughout *Plandemic*, discredited former National Cancer Institute scientist Dr. Judy Mikovits is shown in a series of interview clips (Alba, 2020). Dr. Mikovits makes several demonstrably false or misleading claims about COVID-19, including: 1) coronavirus may have originated from US government research into the flu vaccine, 2) COVID-19 vaccine is being used to push a pro-vaccine agenda led by academia and industry, 3) Dr. Anthony Fauci, director of the US National Institute for Allergy and Infectious Disease, profited from the HIV/AIDS epidemic and suppressed Dr. Mikovits' anti-vaccine research, and 4) face masks activate dormant coronavirus particles implanted through flu vaccination (Alba, 2020; Elliott, 2020; Frenkel et al., 2020; Funke, 2020a). By May 7<sup>th</sup>, 2020 the film was removed from YouTube because it contradicted World Health Organization guidance, a platform policy violation, and nearly all other social media platforms subsequently blocked the film (Lapin, 2020). On August 18<sup>th</sup>, 2020, a 75-minute second half of *Plandemic* was released online but has failed to garner similar attention as the initial release due to pre-emptive actions by online platforms (Funke, 2020b; Spencer et al., 2020).

*Plandemic* delegitimizes, or at least seeks to delegitimize, a vocal and needed advocate for public health risk communication about COVID-19, Dr. Anthony Fauci. In a statement to *Snoopes* following the documentary's release, Dr. Fauci personally refuted Dr. Mikovits' claims (Kasprak, 2020). Nevertheless, delegitimizing visible and trusted public health leaders sows doubt in the federal government's pandemic response as well as the safety and efficacy of an eventual coronavirus vaccine. For example, if Dr. Fauci's character prior to the pandemic is drawn into question, as the documentary suggests, then his post-pandemic conclusions may be unduly scrutinized or even discredited. By extension then, the national shutdown and impacts of quarantine could also be blamed on Dr. Fauci's now-disqualified conclusions. Anti-vaccine activists produced *Plandemic* to increase vaccine hesitancy and decrease vaccination, but their lasting impact may be that it promoted cynicism about measures meant to prevent COVID-19 spread, such as use of face masks and social distancing. Disregarding these measures threatens public health and may only serve to extend the pandemic. Stopping the spread and influence of *Plandemic* – and related misinformation – is in the interest of the public's health.

#### *Plandemic as a case study for social media misinformation*

We investigated tweets about the plandemic conspiracy theory as a case study of COVID-19 social media misinformation, and provide novel and original research that complements media reports (Frenkel et al., 2020). We found that the documentary reduced the online movements' focus on the COVID-19 pandemic as well as vaccine-related conversations. Instead, the film identified specific individuals as the new, personalized focus of anti-pandemic ire. Although conspiracy theories were common and popular both before and after the film's release, post-documentary tweets were particularly focused on personal attacks and vilifying specific public health experts. Our findings support a pro-active, responsive, and multi-pronged effort to fight misinformation, specifically: 1) preemptively addressing misinformation about the COVID-19 vaccine now 2) ongoing monitoring and surveillance of social media platforms for emerging conspiracy theories, 3) censoring, fact-checking, and debunking social media content that contains false information (i.e., misinformation, disinformation, conspiracy theories), and 4) learning about who creates misinformation and conspiracy theories and their motivations. At a minimum, social media platforms must respond to false information, and public health advocates and health educators must also participate through preventive interventions, campaigns, and research.

First, we must preemptively address misinformation about the COVID-19 vaccine. A recent poll reported that half of Americans would get the coronavirus vaccine, while one in five would not get vaccinated (AP/NORC, 2020). Anti-vaccine proponents used *Plandemic* to sow discord with the ultimate

goal of increasing vaccine hesitancy and decreasing vaccination. As the world is now waiting on a vaccine, misinformation such as that contained in *Plandemic* may increase hesitancy towards coronavirus vaccination (Wadman, 2020). In cases where we cannot prevent exposure to misinformation, we need to build health and media literacy skills to prevent misinformation’s *influence* – referred to as primary prevention in public health. Primary prevention of misinformation’s influence – pre-bunking rather than debunking – is necessary and may be accomplished through building literacy skills to discern true versus misleading or false information and develop skepticism towards newly-received information and its sources (Jolley & Douglas, 2017; Lewandowski & Cook, 2020). For example, Roozenbeek and colleagues (2020) created a social media simulation where participants acted as creators of fake news content. Playing the game increased participants’ resistance to misinformation about politics and related conspiracy theories. Within the context of COVID-19, an adapted version of the game could be created with the goal of pre-empting coronavirus vaccine misinformation. Alternatively, another option to preempt misinformation may be to produce a documentary promoting coronavirus vaccination to counter misinformation like *Plandemic*, which cost less than \$2,000 to create (Rottenberg & Perman, 2020). Whether literacy-building content is formatted as a game or viral video, public health advocates and communications must address misinformation about the COVID-19 vaccine now and consider how to effectively use pre-bunking as a mitigation strategy for misinformation’s influence among the general public (Cook et al., 2017; Jolley & Douglas, 2017; Roozenbeek et al., 2020).

Second, there must be robust systems for tracking the emergence of misinformation and conspiracy theories, including the use of both automated and manually-controlled online dashboards (Resnick et al., 2018). Twitter is a uniquely accessible source of social media data because it is open source (i.e., publicly available). While vaccine-related plandemic tweets were limited in scope overall and did not become more prevalent in the wake of the *Plandemic* documentary, the documentary was successful at elevating the profile of the plandemic conspiracy theory in general. Despite a surge in the number of plandemic tweets, where one may expect the average level of engagement (i.e., likes, retweets) to decrease as the amount of content increases, we found that the numbers of likes and retweets were actually unchanged pre- versus post-documentary, suggesting consistent engagement from a reliable network of Twitter users. By harnessing a term/hashtag that was already in use on social media, the film’s promoters had access to pre-existing and captive social networks through which to initiate the spread of the documentary. Online resources are emerging to investigate and track different conspiracy theories and viral misinformation related to COVID-19 (Chen et al., 2020; National Press Foundation, 2020; Nsoesie & Oladeji, 2020). Such tools support journalists as they respond in real-time to developing stories about viral misinformation, and public health researchers must adapt to evolving technologies. For example, Project RCAID (Rapid Collection Analysis Interpretation and Dissemination) is an online misinformation dashboard created through a nonprofit-private partnership (Public Good Projects & Signal Labs, 2020) for tracking emerging coronavirus stories and narratives, as well as new hashtags, on various social media platforms. Through misinformation monitoring and surveillance, the pre-emptive verification, labelling, or even removal of false social media content may be enhanced to further prevent its spread.

Third, censoring, fact-checking, and debunking content that contains false information must take a clear and transparent strategy. Social media and coronavirus-related changes in daily media consumption have created an environment of information overload and increased the use of censoring, labeling, and verifying content by social media platforms (e.g., Twitter) and forum moderators (e.g., Reddit) (Andrews, 2020; Funke, 2020a; Hall Jamieson & Albarracín, 2020). Fact-checkers and platform administrators who verify content may be overwhelmed and unable to respond quickly to emerging misinformation, demanding innovative solutions such as volunteer fact-checkers (Kim & Walker, 2020). In addition, specific pieces of misinformation or conspiracy theories may be directly debunked/fact-checked through providing corrections and alternate explanations, and repeated exposure has been shown to increase the impact of fact-checking (Lewandowsky et al., 2012). Based on our findings, tweets about the plandemic

conspiracy theory spiked post-film but receded towards pre-film volumes – future avenues of research should examine what role media fact-checking, social media platform censorship, and use of warning labels played in this trajectory.

Determining how and when to verify or remove (i.e., censor) social media content is a nuanced debate. On the one hand, removing false content early can prevent the spread of “viral” misinformation from fringe social media communities to the general public and mainstream media outlets. On the other hand, too much censorship and labelling may risk further polarization, particularly among social media platform preferences (Wellemeier, 2020). Furthermore, removing false content from one source may only enhance its spread through other sources as more people share or seek out the removed content (Bellemare et al., 2020). To this end, we found that website links and online media sharing proliferated after the film’s release on May 4<sup>th</sup>, perhaps in response to platforms restricting and removing the film. In the absence of a comprehensive strategy, *Plandemic* may continue to spread through online social media, and potentially traditional media sources such as broadcast television. As recently as July of 2020, *Plandemic* was set to be aired on nearly 200 local US television stations owned by Sinclair Broadcasting Group in a since-abandoned planned broadcast (Farhi, 2020). Sinclair Broadcasting Group is a conservative media conglomerate that operates local television stations across the US, reaching approximately four in ten (39%) US television viewers (A. Chang, 2018). Taken together, our findings and the continued threat posed by *Plandemic* and similar misinformation support a comprehensive and truly multi-media strategy for determining when, how, why, and by whom false content is removed, labelled, or fact-checked.

Fourth, our efforts to address misinformation must understand who posts content and why. Research into cognitive biases, in particular information and confirmation biases, suggests that if someone is convinced of a certain viewpoint and encounters information that refutes their position, then that person would be unlikely to accept new information and may even attack the source of the counter information (Baron et al., 1988; Nickerson, 1998). People who believe conspiracy theories like the plandemic one, may be resistant to fact-checking and easily rationalize new information – both true and false – within the context of existing conspiracy theory beliefs (Ognyanova et al., 2020; Uscinski et al., 2020; Wood et al., 2012). Future research must seek to understand the underlying drivers of anti-science and anti-evidence sentiment by learning about the priorities and health beliefs of people who subscribe to conspiracy theories like plandemic and are influenced by similar misinformation.

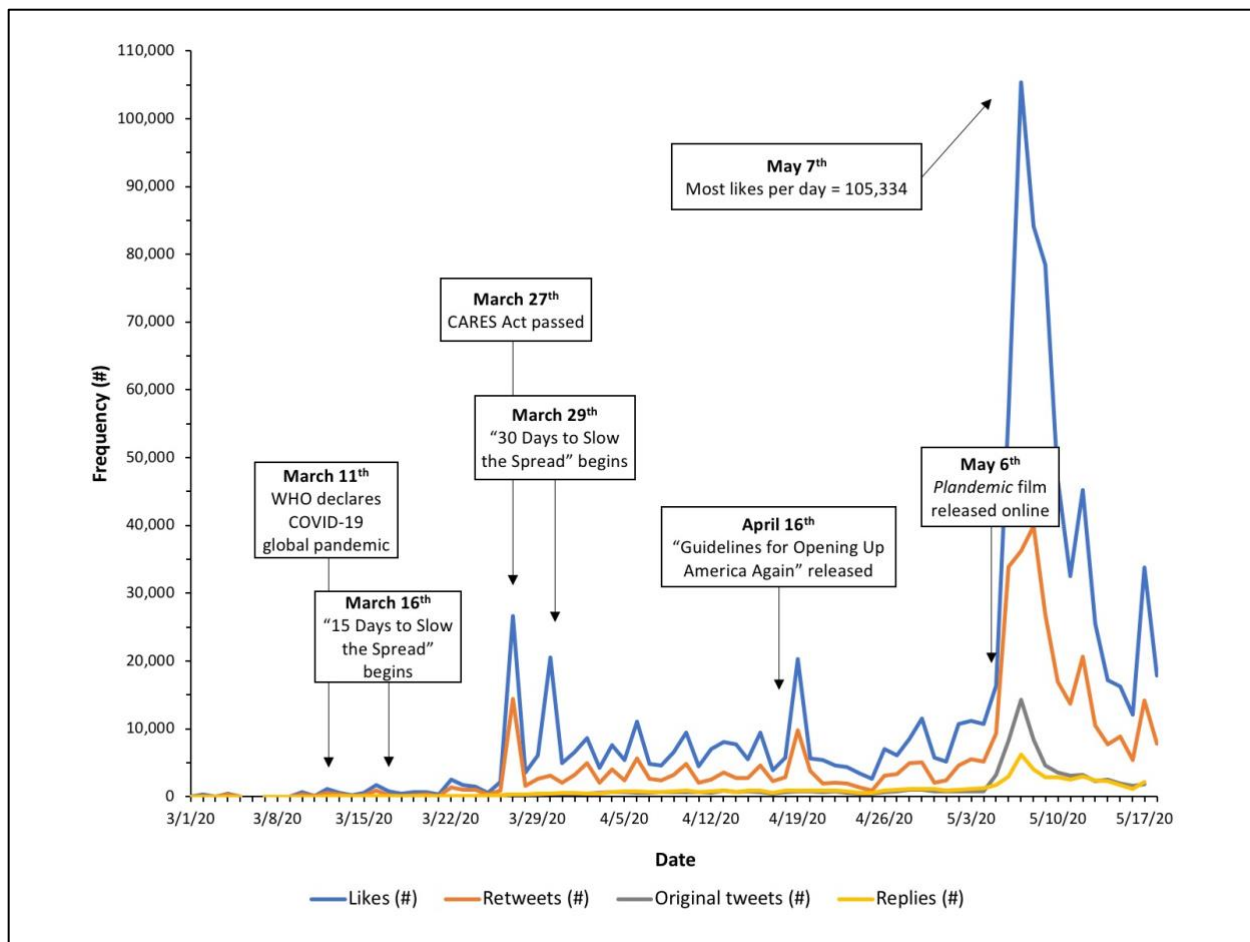
## Findings

Findings are organized by research questions: all three findings address our primary research question (RQ1: *How did the May 2020 release of the documentary Plandemic change ongoing Twitter discourse mentioning the term “plandemic”?*); findings 2 and 3 correspond to research question 2 (RQ2: *What are the characteristics of “plandemic” tweets before versus after the film?*); findings 2 and 3 also correspond to research question 3 (RQ3: *Which tweet characteristics are associated with more likes and retweets before versus after the film?*). Figure 1 presents a timeline of the volume of original tweets, retweets, replies, and likes related to the use of plandemic on Twitter. All results tables are included in Appendix A.

*Finding 1: Twitter discourse about the film spiked and quickly fell off after the Plandemic’s release.*

The film created a surge of new content. More than twice as many tweets were created in the 14 days following the documentary’s release compared to the previous 100 days (see Figure 1). As shown in the examples provided in Table A2 (see Appendix A), the film was first mentioned on May 4<sup>th</sup>, 2020, but the term “plandemic” was used on Twitter as early as January 24<sup>th</sup>, 2020. An average of 1,246 tweets per day mentioned plandemic, and the average number of tweets per day increased more than 10-fold from 612

pre-film to 10,657 post-film. The average number of followers per user more than tripled following the film’s release (see Appendix A, Table A1). No significant change was observed in the average numbers of likes and retweets, suggesting a similar level of user engagement before and after the film across a larger overall volume of plandemic tweets.



**Figure 1. #Plandemic Tweet Volume and Key Events: March 1<sup>st</sup> – May 17<sup>th</sup>, 2020.** The number of likes, retweets, original tweets, and replies are presented for tweets in our dataset (abridged to begin March 1<sup>st</sup>) and annotated with key events for the United States COVID-19 pandemic.

Links and online media sharing proliferated after May 4<sup>th</sup>, perhaps in response to platforms restricting and removing the film. In general, most tweets (58%) were related to online information; nearly half (45%) promoted sharing media, and fewer discussed information censorship (3%) or were about false information (4%) (see Appendix A, Table A3). Percentage of tweets that shared online media nearly tripled from 20% pre-film to 56% post-film, while censorship mentions increased five-fold. Terms used to describe false information nearly doubled after Plandemic (see Appendix A, Table A3). Tweets mentioning online information received more retweets and likes following Plandemic’s release (see Appendix A, Table A3).

*Finding 2: Anti-government and largely anti-liberal political conspiracy theory tweets were common and popular both before and after the film’s release, and tweets were often directed towards conservative media and political figures through mentions and replies.*

A third of tweets (30%) discussed at least one conspiracy theory genre. The most common conspiracy theories were about the deep state government (17%), nefarious cover-ups (13%), anti-vaccination (8%), Bill Gates (6%), and 5G wireless technology (2%) (see Appendix A, Table A3). Tweets about nefarious cover-up conspiracy theories increased in volume post-documentary, whereas other conspiracy theory categories decreased significantly. Politics and government and the media were mentioned less frequently (15% and 2%, respectively) (see Appendix A, Table A3). The three most frequently mentioned political or government officials were current US President Donald Trump (8%), NIAID Director Anthony Fauci (7%), and former US President Barack Obama (2%).

Liberal politics were discussed nearly ten times more often than conservative politics (7% liberal versus <1% conservative, respectively) (see Appendix A, Table A3). Conversely, conservative media and political figures were often targeted through user mentions and replies to recent tweets. Reply tweets were most frequently addressed towards the following users: 1) @RealDonaldTrump, 2) @IngrahamAngle (Fox News host), 3) @RealJamesWoods (conservative American actor; 0.47%), 4) @GregGutfeld (Fox News host), 5) @Mitchellvii (conservative American radio host), 6) @BillGates, and 7) @DrJudyAMikovits (*Plandemic* documentary).

*Finding 3: On Twitter, the film turned attention to certain public figures and the pandemic's impact on freedoms and liberties, but vaccine-related tweets were relatively marginal and unaltered by the film.*

Before the film, the term “plandemic” was associated with a range of conspiracy theories about the “deep state,” Bill Gates, and the pro-vaccine agenda, and President Trump and the political left were most frequently mentioned (see examples tweets in Appendix A, Table A1). After the film was released, the conversation changed decidedly both in terms of the type of content being produced and what content was receiving the most engagement from other Twitter users (see Appendix A, Tables A3 and A4). Tweets in our sample created after the film’s release were less likely to discuss COVID-19 than posts from before (see Appendix A, Table A3). COVID-19’s impact on freedoms and liberties was mentioned significantly more often and received more likes and retweets post-documentary. Anti-vaccine tweets appear to be limited and not changed in the documentary’s wake. Despite an observed decrease in overall tweets about vaccination in our sample, vaccine-related tweets were retweeted and liked more following the documentary’s release (see Appendix A, Tables A3 and A4).

The film polarized Twitter use of plandemic resulting in increased attention towards Barack Obama and Dr. Anthony Fauci. Significant increases in retweets and likes were observed among tweets mentioning former President Barack Obama (770%), NIAID Director Dr. Anthony Fauci (45%), and Bill Gates (42%) (see Appendix A, Tables A3 and A4). Current US President Donald Trump was the most mentioned person in tweets about plandemic, but tweets mentioning Trump decreased significantly after the documentary’s release (see Appendix A, Table A3). Tweets mentioning current President Donald Trump received more retweets before the documentary than afterward (see Appendix A, Table A4). Percentage of tweets discussing mainstream media increased significantly after the film’s release, and tweets about alternate media significantly decreased in volume (see Appendix A, Table A3).

## Methods

A retrospective analysis of tweets mentioning “plandemic” was conducted to address our stated research questions. Our study of original tweets mentioning “plandemic” before and after *Plandemic*’s release provides descriptive and correlational findings that address current infodemiological needs. The term “plandemic” was our search criteria and therefore refers to all eligible analyze tweets (i.e., all tweets contained the keyword “plandemic”). The study dataset comprised 84,884 original tweets mentioning

plandemic from 51,021 Twitter users. Original tweets were liked or shared 1,286,062 times (likes=893,299; retweets=392,763) by other users and could have reached a minimum of 819,112,285 followers.

#### *Data collection and sampling*

Using the “twitter2stata” package in Stata/IC 15.1 software, we downloaded all publicly-accessible tweets (N=545,054) mentioning plandemic between January 24<sup>th</sup> (first mention on Twitter) and May 17<sup>th</sup> (two weeks [14 days] after the film’s May 4<sup>th</sup> online release). Tweets were collected at least once daily depending on the rate limit maximum of 18,000 tweets. We modified code in Stata/IC 15 to collect tweets based on their sequential unique IDs (e.g., 1220727066599010304). Nevertheless, some relevant tweets may not have been captured through this data collection strategy, particularly during period when tweet volume was highest, and this could reduce the transferability of our findings about all plandemic tweets. When the maximum limit was reached, multiple subsequent searches were conducted to reduce the number of tweets that may be missed, such every four hours.

To allow for comparison pre- and post-film, we ended data collection two weeks following the film’s release. The dataset included original tweets (n=84,884), retweets (n=392,964), and tweets that were replies to other tweets (“replies”; n=67,206). Our main analysis of text terms and topics (see example terms in Appendix B, Table B1) includes only original tweets mentioning plandemic because we want to understand unsolicited and unprompted Twitter content. Only English-language terms were used throughout our analysis because Twitter API defaults to English-language tweets. We also did not identify any tweets in languages besides English. We compared all analyses and results before and after the film’s release on May 4<sup>th</sup>, 2020. All statistical analyses and data management were conducted in Stata/IC 15 software.

#### *Data analysis*

To answer RQ1 (*How did the May 2020 release of the documentary Plandemic change ongoing Twitter discourse mentioning the term plandemic?*), Pearson chi-squared tests and unpaired t-tests assessed significant differences in tweet and user metadata characteristics (see Appendix A, Table A1), as well as in text topic categories (see Appendix A, Table A3). We also calculated percent change ( $\Delta$ ) in average engagement (i.e., likes + retweets) and assessed significant differences using unpaired t-tests (see Appendix A, Table A4).

To answer RQ2 (*What are the characteristics of plandemic tweets before versus after the film?*), we described the terms (i.e., hashtags, users, topics) mentioned most frequently in original tweets and conducted a summative content analysis utilizing automated text analysis and topic modeling methods. Detailed information about our content analysis process, including a table of relevant text terms (see Table B1), can be found in Appendix B.

To answer RQ3 (*Which tweet characteristics are associated with more likes and retweets before versus after the film?*), we evaluated differences in proxy measures for reach and popularity. All topic indicators were non-mutually exclusive independent variables. Reach and popularity, the dependent variables, were operationalized by the number of retweets and number of likes, respectively. As both likes and retweets are measures of a tweet’s popularity and potential exposure to Twitter followers and other social networks, we aggregated total likes and retweets for each original tweet, creating a pooled measure of audience engagement (see Appendix A, Table A4).



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**Ethics**

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**Data availability**

All applicable de-identified data and code are available via the Harvard Dataverse repository: <https://doi.org/10.7910/DVN/PN7UPO>.

## Appendix A: Results

Table A1 presents tweet and user metadata characteristics stratified by pre- versus post-film time period and with differences presented using percent (%) change. Table A2 presents selected tweet examples. Table A3 presents text content characteristics stratified by pre- and post-documentary. Table A4 presents changes in average likes and retweets for text content categories, stratified by pre- and post-documentary.

		All Original Tweets (n= 84,884)	Pre- documentary (n= 25,351)	Post- documentary (n= 59,533)	$\Delta$ (% Change; Post – Pre)	p-value
<b>Table A1.</b> Tweet and user metadata characteristics. All tweets (n= 84,884) were collected between January 24 <sup>th</sup> and May 17 <sup>th</sup> , 2020. Characteristics are stratified by pre- versus post-film time period and sum across rows. Percent change between pre- versus post-film release is presented (“ $\Delta$ ”). Tweet characteristics are not mutually exclusive. Chi-squared and t-tests assessed differences pre-versus post-documentary (alpha=0.05).						
<b>Tweet characteristics</b>						
Text content	<i>Question included (%)</i>	14.89	18.83	13.21	-29.85%	<.001
	<i>Link included (%)</i>	79.53	81.37	78.75	-3.22%	<.001
	<i>Statistic included (%)</i>	1.01	1.77	0.69	-61.02%	<.001
	<i>Tweet geotagged (%)</i>	2.01	1.35	2.30	70.37%	<.001
	<i>Included at least one hashtag (%)</i>	46.00	64.22	38.24	-40.45%	<.001
	<i>Mentioned at least one user (%)</i>	29.38	23.47	31.90	35.92%	<.001
	<i>Total hashtags used (mean #) †</i>	3.66	4.11	3.34	-18.73%	<.001
<i>Total users mentioned (mean #) ††</i>	1.40	1.56	1.36	-12.82%	<.001	
Twitter engagement	<i>Received at least one like (%)</i>	42.14	47.21	39.98	-15.31%	<.001
	<i>Received at least one retweet (%)</i>	25.68	32.15	22.93	-28.68%	<.001
	<i>Tweets per day (mean #)</i>	1,245.63	611.72	10,656.86	1642.11%	<.001
	<i>Likes per tweet (mean #)</i>	8.62	9.44	8.27	-12.39%	0.3468
	<i>Retweets per tweet (mean #)</i>	4.05	4.83	3.71	-23.19%	0.0512
<b>User characteristics</b>						
Social network	<i>Followers of user (mean #)</i>	9,653.18	3,519.86	12,266.61	248.50%	<.001
	<i>Followed by user (mean #)</i>	2,477.64	2,591.19	2,429.26	-6.25%	0.1243
† Includes tweets containing at least one hashtag (n=38,722)						
†† Includes tweets mentioning at least one user (n=24,661)						

**Table A2.** Examples of early and popular tweets mentioning “plandemic”. Qualitative descriptions are provided for additional context in second column from left. Text topic categories correspond to tweet text classifications (see Table B1). All tweets were created between January 24<sup>th</sup> and May 17<sup>th</sup>, 2020. Tweets are organized in ascending order by date (specific dates removed for privacy); for comparison, a bold line demarcates pre- versus post-documentary tweets. Retweets and likes represent counts from when the data were pulled and may not reflect current likes/retweets.

Tweet Text*	Text Topic Categories	Retweets (#)	Likes (#)
<i>'Viral Out Break in China May Be Ten Times the Scale of SARS,' An Expert Says. The Main Stream Media tells me that it's dangerous. A #plandemic (a plandemic) 😊 #Coronavirus is man made because it is a depopulation tool 😊 #BiologicalWeapon #ThesePeopleAreSick [ARTICLE LINK]</i>	<ul style="list-style-type: none"> <li>•Mainstream media</li> <li>•Nefarious cover-up</li> <li>•Anti-vaccine</li> <li>•Online media sharing</li> <li>•Health impacts</li> </ul>	1	0
<i>We should start calling the #coronavirus what it is: A PLAndemic. This plan has been in the works for a while now. It was patented many years ago and rehearsed last Fall in 2019. Research "EVENT 201". 🐸🐸</i>	<ul style="list-style-type: none"> <li>•Bill Gates &amp; global elite</li> <li>•Nefarious cover-up</li> </ul>	124 <i>Highest in February</i>	394 <i>Highest in February</i>
<i>I wonder: Why won't the mainstream media interview the tens of thousands of SURVIVORS of #coronavirus ? You would think that they would want to give people hope that there is a 98% chance of survival ... because they care about us so much, am I right? #Plandemic #WeDemandSurvivorStories</i>	<ul style="list-style-type: none"> <li>•Mainstream media</li> <li>•Health impacts</li> </ul>	247	629
<i>BAM. I found something online that ties the Ukraine Impeachment Investigation's Key Witness to the White House Coronavirus Task Force. #GreatAwakeningWorldwide #Plandemic [VIDEO LINK]</i>	<ul style="list-style-type: none"> <li>•Deep state</li> <li>•Political left</li> </ul>	997	1,734
<i>Host Trish Regan is fired by Fox News for exposing the plandemic to remove President Trump from the oval office. [ARTICLE LINK]</i>	<ul style="list-style-type: none"> <li>•Alternate media</li> <li>•Deep state</li> <li>•Presidential Donald Trump</li> <li>•Online media sharing</li> </ul>	7,327 <i>Highest in March</i>	10,272 <i>Highest in March</i>
<i>I will NEVER forget the current time of American history. In the middle of an epic PlanDEMic, Speaker Pelosi awards herself with a raise, gallons of \$13/pint ice cream, and punishes everyone else. She also gave the middle finger to small businesses.</i>	<ul style="list-style-type: none"> <li>•Political left</li> <li>•Economic impact</li> </ul>	5,542 <i>Highest in April</i>	10,723 <i>Highest in April</i>
<i>Please watch and share the 1<sup>st</sup> installment of Plandemic the documentary, featuring Dr. Judy Mikovits @DrJudyAMikovits. Plandemic, the full-length feature documentary movie, is coming summer 2020. [VIDEO LINK 1] [VIDEO LINK 2]</i>	<ul style="list-style-type: none"> <li>•Plandemic film</li> <li>•Online media sharing</li> </ul>	0	3
<i>PLEASE RETWEET! Long after his Presidency ended Barack Obama continues attending secret meetings with the architects of the Plandemic in Silicon Valley: Warren Buffett, Bill Gates, Anthony Fauci, and the World Commission !!! #StopTheSocialist4thReich [VIDEO LINK]</i>	<ul style="list-style-type: none"> <li>•President Barack Obama</li> <li>•Deep state</li> <li>•Bill Gates &amp; global elite</li> <li>•Dr. Anthony Fauci</li> <li>•Political left</li> </ul>	13,338 <i>Most Retweeted Tweet in Study Sample</i>	14,526

	<ul style="list-style-type: none"> <li>•Online media sharing</li> </ul>		
<p><i>Do you want to know what kept me from watching the "Plandemic" online video? It was all of the people flooding social media demanding me to watch the "Plandemic" video. That was the red flag that the film was fake.</i></p>	<ul style="list-style-type: none"> <li>•Plandemic film</li> <li>•False information</li> </ul>	3,663	24,176 <i>Most Liked Tweet in Study Sample</i>
<p><i>FREEDOM!!! Plandemic: The Documentary About A Global Plan To Take Control Of Our Life, Liberty, Health, and Freedom. #WWG1WGA #Freedom #GreatAwakening #PlandemicDocumentary [VIDEO LINK]</i></p>	<ul style="list-style-type: none"> <li>•Plandemic film</li> <li>•Civil liberties impact</li> <li>•Nefarious cover-up</li> <li>•Global elite</li> <li>•Online media sharing</li> </ul>	27	24
<p><i>My mom questioned vaccines but had never looked much into them. It only took ONE video for my mother to be convinced about the corrupt vaccine industry: the Plandemic documentary with Dr. Judy Mikovits about Anthony Fauci!!! Now, my mom cannot stop researching the corrupt vaccine industry !!</i></p>	<ul style="list-style-type: none"> <li>•Plandemic film</li> <li>•Dr. Anthony Fauci</li> <li>•Anti-vaccine</li> </ul>	0	0
<p><i>Hey there Google/YouTube: your censoring of the video "Plandemic" only turned Dr. Judy Mikovits' book into the #1 best-seller now -- this is called Barbra Streisand effect, stupid! #GreatAwakening #WWG1WGA #GreatAwakeningWorldwide #WWG1WGA_WORLDWIDE [ARTICLE LINK]</i></p>	<ul style="list-style-type: none"> <li>•Plandemic film</li> <li>•Censorship</li> <li>•Deep state</li> <li>•Online media sharing</li> </ul>	5	4
<p><i>We should all wear Guy Fawkes masks for face coverings out in public. They will protect just as much against the Hoax Plandemic!! @realDonaldTrump @GOPoversight @mattgaetz @GOPLeader @DrPaulGosar @SenateGOP @Jim_Jordan [GUY FAWKES MASK IMAGE LINK]</i></p>	<ul style="list-style-type: none"> <li>•Anti-containment</li> <li>•Anti-pandemic</li> <li>•Civil liberties impact</li> <li>•Online media sharing</li> <li>•President Donald Trump</li> <li>•Political right</li> </ul>	85	91
<p>*To protect the privacy of the content-creators, example tweets have been modified from their original text.</p>			

**Table A3.** Percentages of major topics mentioned in our sample of tweets. All tweets (n= 84,884) were collected between January 24<sup>th</sup> and May 17<sup>th</sup>, 2020. Topics are organized using broader domains with sub-categories. Percent change between pre- versus post-documentary is presented in the right-most column (“Δ”). Chi-squared and t-tests assessed differences pre-versus post-documentary (alpha=0.05). Characteristics are not mutually exclusive.

Domain	Category	% of All Tweets (n= 84,884)	% of Pre-Documentary Tweets (n= 25,351)	% of Post-Documentary Tweets (n= 59,533)	Δ (% Change; Post – Pre)
<b>COVID-19</b>  42.93%; n=21,902	<i>Plandemic film</i>	10.01	--	14.23	
	<i>Civil liberties impact</i>	13	10.34	14.13	36.65% ***
	<i>Health impact</i>	1.65	2.52	1.28	-49.21% ***
	<i>Anti-containment</i>	3.15	4.9	2.41	-50.82% ***
	<i>Chinese virus</i>	2.46	4.86	1.44	-70.37% ***
	<i>Pro-containment</i>	5.29	7.8	4.22	-45.90% ***
	<i>Economic impact</i>	3.92	6.54	2.81	-57.03% ***
	<i>Neutral pandemic terms</i>	14.4	18.15	12.81	-29.42% ***
	<i>Anti-pandemic terms</i>	6.35	12.06	3.91	-67.58% ***
<b>Conspiracy Theories</b>  28.98%; n=14,788	<i>Nefarious cover-up</i>	13.42	5.49	16.79	205.83% ***
	<i>Deep state</i>	17.05	19.14	16.16	-15.57% ***
	<i>5G Broadband</i>	1.96	4.12	1.04	-74.76% ***
	<i>Anti-vaccine</i>	8.06	11.29	6.69	-40.74% ***
	<i>Bill Gates &amp; global elite</i>	5.94	11.47	3.59	-68.70% ***
<b>Online Information</b>  58.21%; n=29,697	<i>Online media sharing</i>	45.01	20.18	55.58	175.42% ***
	<i>Censorship</i>	3.31	0.82	4.37	432.93% ***
	<i>False information</i>	4.05	2.76	4.6	66.67% ***
<b>Politics &amp; government</b>  15.35%; n=7,832	<i>Dr. Anthony Fauci</i>	7.21	5.89	7.77	31.92% ***
	<i>President Barack Obama</i>	2.03	0.56	2.33	316.07% ***
	<i>Political left</i>	6.88	6.21	7.16	15.30% ***
	<i>Political right</i>	0.74	0.76	0.73	-3.95%
	<i>Health officials and organizations</i>	1.38	2.07	1.09	-47.34% ***
	<i>President Donald Trump</i>	7.54	8.91	6.95	-22.00% ***
<b>Media</b>  2.09%; n=1,065	<i>Mainstream</i>	1.03	0.44	1.29	193.18% ***
	<i>Alternate</i>	1.07	1.35	0.95	-29.63% ***

\*p<.05; \*\* p<.01; \*\*\* p<.001



**Table A4.** Average likes + retweets for “plandemic” tweets by topic category. All tweets (n= 84,884) were collected between January 24<sup>th</sup> and May 17<sup>th</sup>, 2020. Percent change ( $\Delta$ ) is presented in column 7. Unpaired t-tests assessed differences in mean likes + retweets before versus after Plandemic’s release.

		Pre-Documentary		Post- Documentary		Comparison	
		Likes + Retweets (Mean #)	SD	Likes + Retweets (Mean #)	SD	$\Delta$ (% Change; Post – Pre)	p-value
	All plandemic tweets	14.27	1.45	11.98	0.95	-16.0%	0.1876
<b>COVID-19</b>  42.93%; n=21,902	Health impact	9.26	2.09	31.49	19.79	240.0%	0.9174
	Pro-containment	12.24	2.77	22.42	6.81	83.2%	0.0697
	Anti-containment	16.47	4.89	25.85	10.21	56.9%	<.001
	Civil liberties impact	11.14	1.23	13.02	1.98	16.8%	<.001
	Chinese virus	17.15	4.01	20.01	9.54	16.7%	0.1645
	Anti-pandemic terms	7.45	0.81	7.86	1.27	5.4%	<.001
	Neutral pandemic terms	12.43	1.72	11.39	1.63	-8.4%	<.001
	Economic impact	31.04	10.30	14.42	2.46	-53.5%	0.0009
	Plandemic film	131.75	103.91	12.66	1.97	-90.4%	0.0063
<b>Conspiracy Theories</b>  28.98%; n=14,788	Bill Gates & global elite	10.61	1.24	15.06	3.81	42.0%	<.001
	Anti-vaccine	11.16	1.56	14.35	2.87	28.6%	0.0592
	Deep state	8.22	0.98	7.70	0.88	-6.3%	<.001
	5G Broadband	8.38	1.80	7.07	2.05	-15.7%	<.001
	Nefarious cover-up	12.24	2.62	7.20	2.86	-41.2%	<.001
<b>Online Information</b>  58.21%; n=29,697	False information	7.61	1.53	17.25	3.76	126.6%	0.0064
	Online media sharing	5.80	0.83	9.82	1.11	69.4%	0.001
	Censorship	11.49	5.42	13.81	2.40	20.2%	0.0006
<b>Politics &amp; government</b>  15.35%; n=7,832	President Barack Obama	3.39	0.98	29.54	20.23	770.3%	0.0586
	Health officials and organizations	9.87	1.84	15.05	7.78	52.5%	0.0001
	Dr. Anthony Fauci	14.30	2.44	20.72	6.58	44.9%	<.001
	Political right	5.15	1.16	7.35	3.47	42.9%	0.0008
	Political left	19.97	4.01	19.05	6.91	-4.6%	<.001
	President Donald Trump	13.16	2.09	9.62	1.88	-26.9%	<.001
<b>Media</b>  2.09%; n=1,065	Alternate	8.18	2.00	6.62	2.76	-19.1%	<.001
	Mainstream	7.05	3.53	3.42	0.72	-51.5%	<.001

## Appendix B: Data analysis

As defined by Hsieh and Shannon (2005), summative content analysis “involves counting and comparisons, usually of keywords or content, followed by the interpretation of the underlying context”. Our content analysis was facilitated by NVivo 12 software (QSR International). In NVivo, automated word frequency queries generated lists of unique hashtags (e.g., #FireFauci), Twitter handles (e.g., @NYGovCuomo), and other text terms (e.g., “lockdown”, “Chinese virus”). The lists contained rows of the most frequent terms mentioned within the entire corpus of the tweet text – up to 280 characters containing numbers, letters, special characters, or emojis. Terms were organized into groups based on their similarity and named according to topic category. As shown in Table B1, topic categories were further grouped under broad domains and were not mutually exclusive. Separate indicator variables (0=no; 1=yes), 27 variables in total, were created for each topical code and descriptive statistics were generated (see Tables 1 and 2). For example, tweets were coded as mentioning US President Trump if any of the following terms were included within the tweet text: “@realdonaldtrump”, “@potus”, “@whitehouse”, “trump”, “#maga”. Confirmatory factor analysis was conducted to assess the robustness (i.e., fit) of our topic modelling, finding that our topic categories (n=27 total) accounted for 98.5% of variance in the dataset across 12 key factors (Brown, 2015). To briefly characterize reply tweets, we identified top recipient users – in other words, users who were most often targeted with tweets mentioning plandemic.

**Table B1.** Examples of topic category key text terms. Analyses were not case sensitive; capitalization is used below when appropriate or for emphasis.

Domains	Category	Term criteria examples
COVID-19	<i>Anti-pandemic terms</i>	“scamdemic”, “shamdemic”, “plannedemic”, “mandemic”, “covidiot”
	<i>Neutral pandemic terms</i>	“#covid19”, “covid-19”, “#coronavirus”, “coronavirus”, “pandemic”
	<i>Pro-containment</i>	“#socialdistancing”, “wearmask”, “#stayhome”, “#donotopen”
	<i>Anti-containment</i>	“#endthelockdowns”, “#getbacktowork”, “#openamerica”, “liberate”
	<i>Chinese virus</i>	“chinese virus”, “china virus”, “wuhan virus”, “wu flu”, “kungflu”
	<i>Economic impact</i>	“pension”, “bailout”, “strife”, “markets”, “commerce”, “business”
	<i>Civil liberties impact</i>	“freedom”, “liberty”, “rights”, “constitution”, “#wethepeople”
	<i>Health impact</i>	“people are dying”, “[#] people die”, “hospitalized”, “health”, “risk
	<i>Plandemic film</i>	“#plandemicdocumentary”, “#plandemicfilm”, “@drjudyamikovits”
Conspiracy Theories	<i>Bill Gates &amp; global elite</i>	“bill gates”, “#arrestbillgates”, “#event201”, “#id2020”
	<i>Nefarious cover-up</i>	“#coverup”, “cover up”, “secret”, “planned”, “set up”, “hidden”
	<i>Deep state</i>	“qanon”, “qarmy”, “#newworldorder”, “deep state”, “#wwg1wga”
	<i>5G Broadband</i>	“5g”, “ggggg”, “fiveg”, “5 g”, “5-g”
	<i>Anti-Vaccine</i>	“antivax”, “vacscene”, “vaxxed”, “wakefield”, “robert f. kennedy jr.”
Online Information	<i>False information</i>	“misinformation”, “false claims”, “false information”, “debunked”
	<i>Censorship</i>	“censor”, “free speech”, “first amendment”, “removed”, “blocked”
	<i>Online media sharing</i>	“youtube”, “reddit”, “bitchute”, “vimeo”, “video link”, “watch here”
Politics & government	<i>Political left</i>	“Pelosi”, “Cuomo”, “democrat”, “#planDEMic”, “#panDEMic”
	<i>Political right</i>	“@senatemajldr”, “McConnell”, “gop”, “republican”, “conservatives”
	<i>President Donald Trump</i>	“@realdonaldtrump”, “@potus”, “@whitehouse”, “trump”, “#maga”
	<i>President Barack Obama</i>	“@barackobama”, “Barack Obama”, “Obamagate”
	<i>Dr. Anthony Fauci</i>	“#firefauci”, “fauci”, “fire fauci”, “#faucifraud”, “fauci fraud”
	<i>Health officials and organizations</i>	“CDC”, “FDA”, “NIH”, “WHO”, “Birn”, “Surgeon General”
Media	<i>Alternate</i>	“Ingraham”, “Hannity”, “Limbaugh”, “Breitbart”, “Infowars”
	<i>Mainstream</i>	“@msnbc”, “@cnn”, “@nytimes”, “@washingtonpost”, “@abcnews”