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Citation impact and social media visibility of Great Barrington and John Snow signatories for COVID-19 strategy

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ABSTRACT

Objective The Great Barrington Declaration (GBD) and the John Snow Memorandum (JSM), each signed by numerous scientists, have proposed hotly debated strategies for handling the COVID-19 pandemic. The current analysis aimed to examine whether the prevailing narrative that GBD is a minority view among experts is true.

Methods The citation impact and social media presence of the key GBD and JSM signatories was assessed. Citation data were obtained from Scopus using a previously validated composite citation indicator that incorporated also coauthorship and author order and ranking was against all authors in the same Science-Metrix scientific field with at least five full papers. Random samples of scientists from the longer lists of signatories were also assessed. The number of Twitter followers for all key signatories was also tracked.

Results Among the 47 key GBD signatories, 20, 19 and 21, respectively, were top-cited authors for career impact, recent single-year (2019) impact or either. For comparison, among the 34 key JSM signatories, 11, 14 and 15, respectively, were top cited. Key signatories represented 30 different scientific fields (9 represented in both documents, 17 only in GBD and 4 only in JSM). In a random sample of n=30 scientists among the longer lists of signatories, five in GBD and three in JSM were top cited. By April 2021, only 19/47 key GBD signatories had personal Twitter accounts versus 34/34 of key JSM signatories; 3 key GBD signatories versus 10 key JSM signatories had >50 000 Twitter followers and extraordinary Kardashian K-indices (363–2569). By November 2021, four key GBD signatories versus 13 key JSM signatories had >50 000 Twitter followers.

Conclusions Both GBD and JSM include many stellar scientists, but JSM has far more powerful social media presence and this may have shaped the impression that it is the dominant narrative.

Strengths and limitations of this study

► Citation impact metrics and Twitter followers can be measured with relatively high accuracy.
► The analysis focused primarily on the key signatories.
► Both citation indices and Twitter followers have limitations in face validity and construct validity as measures of impact.

released with a short time difference in the fall of 2020. Each of them had a core team of original signatories and over time signatures were collected for many thousands of additional scientists, physicians and (in the case of GBD) also citizens.[4] A careful inspection is necessary to understand the differences (but also potential common points) of the two strategies.[4 5] The communication of these strategies to the wider public through media and social media has often created confusion and tension. The communication includes what endorsing scientists state and how opponents describe the opposite strategy. Oversimplification, use of strawman arguments, and allusions of conflicts, political endorsements and ad hominem attacks can create an explosive landscape.[4–9]

Briefly, GBD is focused on targeted protection of high-risk individuals, while JSM considers that such a strategy may not be achievable. Much tension[5–9] surrounds also the concept of herd immunity, where GBD declares that herd immunity is unavoidable eventually (much like gravity is unavoidable), while JSM stresses that aiming for herd immunity through natural infection is unethical. JSM proponents often accuse GBD proponents as urging the population to be infected, while GBD signatories deny this accusation. The two schools also tend to differ in terms of their approach towards lockdowns, seen in a far more negative light in GBD than in JSM.

It is often stated in social media and media, by JSM proponents in particular, that JSM is by far the dominant strategy and that very few

INTRODUCTION

The optimal approach to the COVID-19 pandemic has been an issue of major debate. Scientists have expressed different perspectives and many of them have also been organised to sign documents that outline overarching strategies. Two major schools of thought are represented by the Great Barrington Declaration (GBD)[1] and the John Snow Memorandum (JSM)[2 3] that were

scientists with strong credentials endorse GBD.[6–9] GBD proponents are often characterised as fringe, arrogant and wrong by their opponents.[6–9] However, are these views justified based on objective evidence on scientific impact or they reflect mostly perceptions created by social media and their uptake also by media?

Here, an analysis is being performed to try to evaluate the scientific impact and the social media visibility of the key signatories who have led the two strategies. Scientific impact is very difficult to evaluate in all its dimensions and no single number exists that can measure scientific excellence and scholarship. However, one can use citation metrics to objectively quantify the impact of a scientist's work in terms of how often it is used in the scientific literature. Adjustments for coauthorship patterns, relative contributions and scientific field need to be accounted for.[10] Concurrently, an additional analysis evaluated the social media visibility of signatories, as denoted by Twitter followers.

METHODS
Documents and signatories
The two documents were retrieved online.[1–3] For the main analysis, the 47 original key signatories of the GBD who were listed on its original release online, and the 34 original key signatories who authored the first release of the JSM in a correspondence item published in the *Lancet*[3] were considered for in-depth citation analysis.

The two documents have been signed by many more signatories. As of 2 April 2021, the GBD site[1] listed the following signature counts: 764 172 concerned citizens, 13 796 medical and public health scientists and 41 895 medical practitioners. However, detailed data on names and affiliations were provided only on 443 medical and public health scientists. As of 25 November 2021, signature counts included 811 461 concerned citizens, 15 019 medical and public health scientists and 44 541 medical practitioners. As of 2 April 2021, the JSM site[2] listed 3600 names of signatories (expanded to 4200 as of 25 November 2021). The sets of 443 and 3600 names included also the original 47 and 34 key signatories, respectively. A random set of 30 names was selected from the 443 GBD names and from the 443 first-listed JSM names on 2 April 2021, acknowledging that the earlier listed names may be more likely to include highly cited, prominent scientists.

Citation data
Citation analyses used data on a validated composite citation indicator that considers six citation indicators (total citations, Hirsch H-index, coauthorship-adjusted Hm-index, total citations to single-authored papers, total citations to single or first-authored papers, total citations to single, first or last-authored papers).[10–12] Existing databases were used that contain all authors who are in the top 2% of their scientific field based on career-long impact until the end of 2019 and based on impact in a recent single year (2019).[12] Given that field assignment

is not perfect, scientists who are in the top 100 000 in the composite citation indicator across all scientists across all science are also included, regardless of whether they reach the top 2% in their specific field. Data were available including and excluding self-citations, as previously described,[11 12] and the latter are presented in the results, unless otherwise specified. The databases are compiled based on Scopus information on all authors who have at least five full papers (articles, reviews, conference papers) in their career (~8 million authors). Science is divided in 174 scientific fields according to the Science-Metrix classification that capitalises on the subject matter and journal venues where articles appear.[13]

Twitter information
For the 43 and 34 original key signatories, their names were searched on Google to identify personal Twitter accounts. Only accounts listed under their name were eligible, excluding group or institutional accounts from groups/institutions where they belonged or which they may have led. The number of followers of eligible Twitter accounts as of 2 April 2021 was recorded and an updated search was performed on 25 November 2021.

Kardashian index calculations
The Kardashian K-index[14] is providing an impression on whether the Twitter footprint of a scientist is disproportionately high compared with the footprint of his/her citation impact. It is calculated as the ratio of Twitter followers divided by $43.3C^{0.32}$, where C is the total citations received in one's career. The original publication[14] defining the index used citations from Google Scholar. However, given that many signatories did not have Google Scholar pages and Google Scholar citations may be more erratic, Scopus citations (including self-citations) as of 2 April 2021 were used instead. Scopus citation counts may be slightly or modestly lower than Google Scholar citations, and this may lead to slightly higher K-index estimates, but the difference is probably small.

Patient and public involvement
There was no patient or public involvement in the study. No patients were evaluated in the study.

RESULTS
Top-cited scientists among the key GBD and JSM signatories
Among the 47 original key signatories of GBD, 20, 19 and 21, respectively, were among the top-cited authors for their career impact, their recent single-year (2019) impact or either. Among the 34 original key signatories of JSM, 11, 14 and 15, respectively, were among the top-cited authors for their career impact, their recent single year (2019) or either. The percentage of top-cited scientists is modestly higher for GBD than for JSM, but the difference is not beyond chance (p>0.10 for all three definitions).

Table 1 Top-cited scientists among the key signatories of the Great Barrington Declaration (GBD) and John Snow Memorandum (JSM)

Scientists	Primary field	Secondary field	Single-year rank (2019)	Career-long rank	Scientists in same field
GBD					
Walker, Alexander M	Pharmacology & Pharmacy	Epidemiology	353	78	94611
Dalgleish, Angus G	Oncology & Carcinogenesis	Immunology	2632	922	230678
Brookes, Anthony J	Genetics & Heredity	Developmental Biology	207	211	32641
Janvier, Annie	Pediatrics	Applied Ethics	314	1065	49820
Kotchoubey, Boris	Neurology & Neurosurgery	Experimental Psychology	5446	5595	227881
Meissner, H Cody	Pediatrics	Microbiology	221	263	49820
Katz, D L	Public Health	Nutrition & Dietetics	349	374	48533
Livermore, David M	Microbiology	General & Internal Medicine	17	6	134369
Shahar, Eyal	Cardiovascular System & Hematology	Neurology & Neurosurgery	945	1203	152312
Kampf, Günter	Epidemiology	Microbiology	102	183	9540
Colhoun, Helen M	Endocrinology & Metabolism	Cardiovascular System & Hematology	311	420	69094
Ludvigsson, Jonas F	Gastroenterology & Hepatology	General & Internal Medicine	59	326	76367
Ratcliffe, Matthew	Philosophy	Experimental Psychology	141	*	7775
Levitt, Michael	Biophysics	Bioinformatics	19	6	18401
Hulme, Mike	Meteorology & Atmospheric Sciences	Geography	52	45	54940
Majumder, Partha P	Genetics & Heredity	Evolutionary Biology	544	412	32641
McKeigue, Paul	Endocrinology & Metabolism	Genetics & Heredity	1170	326	69094
Wood, Simon N	Statistics & Probability	Ecology	9	29	16942
Bhattacharya, Jay	Health Policy & Services	General & Internal Medicine	281	*	16521
Kulldorff, Martin	Statistics & Probability	Public Health	58	34	16942
Friedman, Eitan	Oncology & Carcinogenesis	Genetics & Heredity	*	1974	230678
JSM					
Bogaert, Debby	Microbiology	Respiratory System	1812	*	134369
Dowd, Jennifer B	Epidemiology	Public Health	141	*	9540
Goldman, Lynn R	Toxicology	Epidemiology	*	531	45124
Greenhalgh, Trisha	Health Policy & Services	General & Internal Medicine	171	22	106795
Hanage, William P	Microbiology	Developmental Biology	628	1236	134369
Kellam, Paul	Virology	Microbiology	1075	553	58416
Krammer, Florian	Virology	Immunology	21	708	58416
Lipsitch, Marc	Microbiology	Epidemiology	189	172	134369
McKee, Martin	Public Health	General & Internal Medicine	22	19	48533
Rutter, Harry	Public Health	Endocrinology & Metabolism	811	*	48533
Smith, Tara C	Microbiology	Epidemiology	1595	*	134369
Sridhar, Devi	General & Internal Medicine	Developmental Biology	1321	1825	106795
Swanton, Charles	Oncology & Carcinogenesis	Developmental Biology	47	445	230678
Walensky, Rochelle P	Virology	Microbiology	436	670	58416
Yamey, Gavin	General & Internal Medicine	Tropical Medicine	860	766	106795

*Not in the top 2% of the field or top 100 000 across all scientists and all science for the specific time frame.

Table 1 shows the 36 top-cited scientists from the key signatories of the two documents along with their primary and secondary scientific fields and their ranking among all scientists in their primary scientific field. As shown, when both the primary and secondary fields were considered, both documents had top-cited signatories representing nine fields (Developmental Biology, Endocrinology & Metabolism, Epidemiology, General & Internal Medicine,

Health Policy & Services, Immunology, Microbiology, Oncology & Carcinogenesis, Public Health). Conversely 17 fields were represented only by key GBD signatories (Applied Ethics, Bioinformatics, Biophysics, Cardiovascular System & Hematology, Ecology, Evolutionary Biology, Experimental Psychology, Gastroenterology & Hepatology, Genetics & Heredity, Geography, Meteorology & Atmospheric Sciences, Neurology, Nutrition & Dietetics, Pediatrics, Pharmacology & Pharmacy, Philosophy, Statistics & Probability) and four fields were represented only by key JSM signatories (Respiratory System, Tropical Medicine, Toxicology, Virology).

Random samples of scientists from the longer list of signatories

In a random sample of n=30 scientists among the longer list of GBD signatories, five were included in the databases of top-cited authors (in career-long and/or recent single-year citation impact), while this was true for n=3 of 30 JSM controls, a difference not beyond chance (p>0.10). These sampled scientists included three of the key signatories (Helen Colhoun and Michael Levitt (GBD) and Martin McKee (JSM)), and five additional ones (Dusko Ilic (Developmental Biology, Biochemistry & Molecular Biology), Michael Jensen (Endocrinology & Metabolism, General & Internal Medicine), Guy Hutton (Tropical Medicine, Health Policy & Services) in GBD; David Schwappach (General & Internal Medicine, Health Policy & Services) and Jose M Martin-Moreno (General & Internal Medicine, Nutrition & Dietetics) in JSM).
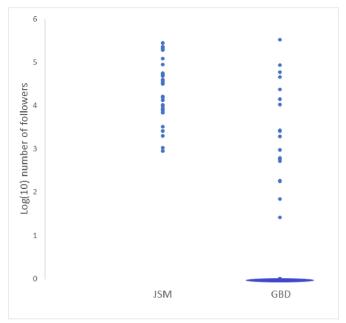
Excluding the key signatories, the proportions were 3/26 and 2/27 (p>0.10). The original key signatories were far more likely to include top-cited scientists in the GBD list (21/47 vs 3/26, p=0.004) and the same was true also for the JSM list (15/34 vs 2/27, p=0.002).

Personal Twitter accounts

As of 2 April 2021, only 19/47 key GBD signatories had a retrievable personal Twitter account, while every single one of the 34 key signatories of JSM had a personal Twitter account (p<0.001). The median number of followers of the 34 JSM scientists was much larger than the median number of followers of the 47 GBD scientists (31 600 vs 0, p<0.001, figure 1).

Only 4/47 GBD signatories versus 17/34 JSM signatories had over 30 000 Twitter followers (3/47 vs 10/34 for signatories with over 50 000 Twitter followers). Twitter and citation data, and inferred Kardashian K-indices for the scientists with >50 000 followers appear in table 2. The values of K-index in these scientists were extraordinarily high (363–2569).

An updated search for Twitter accounts and followers on 25 November 2021 found that 22/47 key GBD signatories versus 34/34 key JSM signatories had a retrievable Twitter account (p<0.001). The median number of followers was 0 vs 34 600 (p<0.001). The number of key signatories with >50 000 followers was 13 vs 4.

Figure 1 Number of Twitter followers of John Snow Memorandum (JSM) and Great Barrington Declaration (GBD) key signatories in April 2021. Twenty-eight of the 47 GBD signatories had no identified personal Twitter accounts.

DISCUSSION

An analysis of citation and social media impact of GBD and JSM signatories shows that both documents have been signed by many leading stellar scientists with very high citation impact in the scientific literature. Random sample data on the longer list of signatories suggest that, expectedly, the longer lists are less thickly populated with extremely highly cited scientists. The total number of top-cited scientists cannot be compared for the two documents because the GBD site does not provide details on all the signatories and signatures are still verified and vetted. Thus, it is unclear whether the much larger total number of signatures in GBD would also translate to a substantially larger total number of top-cited scientists endorsing it as compared with JSM. Regardless, GBD is clearly not a fringe minority report compared with JSM, as many social media and media allude.[6–9] GBD may be a more commonly espoused narrative than the JSM narrative among most cited scientists. Acknowledging uncertainty given the fragmentary nature of the presented names of signatories, it is safe to conclude that both documents have been endorsed by many scientists who are very influential in the scientific literature.

Conversely, the two cohorts of key signatories have a stark difference in Twitter follower counts. The majority of key GBD signatories have no personal Twitter account that could be readily identified. While it is possible that some accounts might have been missed (eg, if not directly named after the individual scientist's names), the difference is so major that it is very unlikely to be a data retrieval artefact. Even among those GBD signatories who do have Twitter accounts, very few have a high number of followers. The key JSM signatories have a

Table 2 Key signatories of John Snow Memorandum (JSM) or Great Barrington Declaration (GBD) with over 50 000 Twitter followers as of April 2021: citation impact, H-index and K-index

Scientist	Twitter followers (April 2021)*	Twitter followers (November 2021)*	Citations	H-index	K-index
Deepti Gurdasani (JSM)	50 400	103 900	5799	19	454
Angela Rasmussen (JSM)	210 800	283 900	1378	18	1931
Dominic Pimenta (JSM)	53 900	58 300	37	2	997
Trisha Greenhalgh (JSM)	121 700	150 000	28 003	81	689
Nisreen Alwan (JSM)	50 700	69 100	1059	19	456
Emma Hodcroft (JSM)	56 900	65 700	577	13	578
Florian Krammer (JSM)	192 600	232 300	11 288	61	1194
Marc Lipsitch (JSM)	226 900	235 800	23 565	82	1279
Devi Sridhar (JSM)	281 100	310 100	2720	23	2380
Rochelle Walensky (JSM)	90 000	345 800	10 561	54	580
Karol Sikora (GBD)	330 300	331 800	4401	30	2569
Michael Levitt (GBD)	86 900	105 300	40 731	106	451
Martin Kulldorff (GBD)	58 800	171 300	14 081	62	363

Only scientists with >50 000 Twitter followers as of April 2021 are shown in the table. As explained in the asterisk footnote below, by November 2021, there were three more John Snow Memorandum key signatories (Isabella Eckerle, Zoe Hyde, Viola Priesemann) who had increased their Twitter followers to >50 000 and one more Great Barrington Declaration signatory (Jay Bhattacharya) who had acquired a Twitter account in the meanwhile and had also exceeded 50 000 Twitter followers. H-indices in November 2021 were 26 for Isabella Eckerle, 22 for Zoe Hyde, 21 for Viola Priesemann and 37 for Jay Bhattacharya.

*Twitter followers for other key signatories in April 2021 and (in parenthesis) in November 2021: Rochelle A. Burgess 3281 (4504), Simon Ashworth 8246 (9124), Rupert Beale 15 500 (19 200), Nahid Bhadelia 33 700 (39 400), Debby Bogaert 2574 (3030), Jenn Dowd 6933 (7221), Isabella Eckerle 48 800 (61 200), Lynn R Goldman 909 (922), Adam Hamdy 10 100 (12 000), William Hanage 39 500 (48 700), Zoe Hyde 39 100 (55 300), Paul Kellam 1069 (1107), Michelle Kelly-Irving 10 200 (10 500), Alan McNally 16 300 (19 300), Martin McKee 33 800 (40 300), Ali Nouri 31 600 (34 600), Viola Priesemann 37 700 (54 900), Harry Rutter 8714 (8859), Joshua Silver 13 300 (15 800), Charles Swanton 7724 (8721), Gavin Yamey 10 200 (26 100), Hisham Ziauddeen 2025 (9795) for JSM; and Andrius Kavaliunas 182 (3479), Ariel Munitz 952 (1013), David Katz 46 000 (46 000), Eyal Shahar 2619 (6152), Gabriela Gomes 10 500 (14 400), Gerhard Krönke 69 (117), Jonas Ludvigsson 2693 (7140), Lisa White 586 (642), Matthew Strauss 14 200 (22 700), Rajiv Bhatia 187 (1525), Salmaan Keshavjee 1955 (2213), Simon Thornley 520 (1207), Sylvia Fogel 614 (3405), Udi Qimron 2695 (4374), Yaz Gulnur Muradoglu 26 (39) for GBD. No personal Twitter accounts were found for the remaining GBD signatories in April 2021, but three of them had detectable Twitter accounts in searches done in November 2021 (Ellen Townsend 18 400 followers, Stephen Bremner 15 followers, Jay Bhattacharya 80 800 followers).

very large number of followers in highly active personal Twitter accounts. The most visible Twitter owners include some of the most cited scientists in the analysed cohorts (Trisha Greenhalgh, Marc Lipsitch, Florian Krammer, Rochelle Walensky, Michael Levitt, Martin Kulldorff, Jay Bhattacharya) and others who have little or no impact in the scientific literature, but are highly remarkable and laudable for their enthusiastic activism (eg, Dominic Pimenta).

Previous work that introduced the Kardashian K-index stated that K-index values above 5 suggest an overemphasis of social media versus scientific literature presence and called such researchers 'Science Kardashians'.[14] This characterisation has not caught up with evolutions in the last few years. Many signatories, especially of JSM, have extraordinarily high K-index, with values in the hundreds and thousands. However, one should account that the volume of Twitter users and followers has increased markedly since the K-index was first proposed, even before the COVID-19 pandemic and even for specialists in disciplines that are not very likely to attract massive social media interest (eg, urology).[15] As COVID-19 has attracted

tremendous social media attention, Kardashian K-indices are skyrocketing. While no past data were available for the number of followers of the analysed scientists pre-COVID, anecdotal experience suggests that many, if not most, saw their followers increase tremendously during the pandemic. Substantial increases were documented even in the short 7-month interval between April and November 2021.

The massive advent of social media contributes to a rampant infodemic[16–18] with massive misinformation circulating. If knowledgeable scientists can have strong social media presence, massively communicating accurate information to followers, the effect may be highly beneficial. Conversely, if scientists themselves are affected by the same problems (misinformation, animosity, loss of decorum and disinhibition, among others)[19 20] when they communicate in social media, the consequences may be negative.

The current analysis has several limitations. The analysis focused primarily on the key signatories and only a small sample of the other signatories from the longer lists was perused. More importantly, both citation indices

and Twitter followers have limitations in face validity and construct validity as measures of impact. A lesser concern is also that both can have errors of measurement, as discussed below. The most important caveat is that scientific impact is difficult to capture fully with any quantitative metrics.

Specifically, citation indices do not capture necessarily all aspects of scholarship.[21] The standardised, validated composite index used here overcomes many of the limitations of crude citation counting, but it is still not perfect. For detailed description of the methods (and their validation) involved in selecting the top-cited scientists across disciplines, one is referred to the background work done to generate the lists of top-cited scientists.[10–12] Precision and recall (author disambiguation in assigning papers) are not perfect in Scopus, and some authors may have underestimated or overestimated citation metrics, but large errors are very uncommon.[22] Publications in Scopus author profiles have 98.1% average precision (ratio of publications correctly assigned to an author) and 94.4% average recall (ratio of an author's papers captured compared with a gold set).[22] The precision for citation linking in Scopus is measured at 99.9% and the recall is 98.3%.[22] Regardless, of the high technical accuracy of these citation data, many scientists who are not included in the lists of top-cited scientists may be at least as outstanding as those who are included, and many dimensions of scholarship, social responsibility and broader impact may be missed by citation indices.[23]

Twitter follower counts are practically impossible to see as measures of excellence in the absence of context. Social media impact may not necessarily be positive, and massive misinformation and despicable behaviour may still generate huge follower lists. Personal Twitter accounts are easy to match against a specific person, provided that the identity of that person can be discerned in Twitter. One cannot exclude the possibility that some of the people for which no Twitter account could be identified may have a pseudonymous Twitter account that hides their true identity. However, in this case, they are not using their personal credentials and overall expertise profile to support the credibility and validity of their Twitter content. Moreover, some academics or researchers may not have personal Twitter accounts, but the centre, institute or other organisation they work in may have some social media presence. The current analysis did not aim to capture these Twitter accounts, since, by definition, they are not personal accounts, but serve a very different role.

Acknowledging these caveats, the data suggest that the massive superiority of JSM over GBD in terms of Twitter firepower may have helped shape the narrative that it is the dominant strategy pursued by a vast majority of knowledgeable scientists. This narrative is clearly contradicted by the citation data. The Twitter superiority may also cause, and/or reinforce also superiority in news coverage. In a darker vein, it may also be responsible for some bad publicity that GBD has received, for example, as evidenced by plain Google

searches online or searches in Wikipedia pages for GBD, its key signatories or even for other scientists who may espouse some GBD features, for example, scepticism regarding the risk-benefit of prolonged lockdowns. Smearing, even vandalisation, is prominent for many such Wikipedia pages or other social media and media coverage of these scientists. This creates a situation where scientific debate becomes vitriolic, and censoring (including self-censoring) may become prominent. Perusal of the Twitter content of JSM signatories and their op-eds suggests that some may have sadly contributed to GBD vilification.[24]

A major point of attack has been alleged conflicts of interest. However, GBD leaders have repeatedly denied conflicts of interest (see also the site of GBD[1]). Key JSM signatories appropriately and laudably disclosed upfront all potential conflicts of interest in their original letter publication in the *Lancet*; the long list is available in public.[3] Based on this list, it is possible that JSM leaders have more conflicts than GBD leaders, but the social media superiority of JSM controls also the narrative surrounding conflicts. A similar vitriolic attack has been launched against the American Institute of Economic Research that offered the venue for hosting the launch of GBD.[24] Experimental studies show that mentioning conflicts may have the same degree of negative impact as attacks on the empirical basis of the science claims; allegations of conflict of interest are as influential as allegations of outright fraud, when the value of scientific evidence is appraised.[25] Non-scientists' trust is eroded by allusions of conflicts of interest, while it is not affected much by perception of scientific (in)competence (which is also impossible for a non-expert to appraise).[25 26] In good faith, reporting of potential conflicts of interest should be encouraged and transparency maximised. However, spurious allegations of hidden agendas and conflicts should not become a weapon for invalidating one or the other document. While exceptions may exist, probably the vast majority of scientists who signed either document simply had good intentions towards helping in a major crisis.

Number of signatures and/or scientific or other impact and visibility of the signatories does not prove that a document is correct. While such petitions are becoming increasingly common in science,[27] it is erroneous to imagine that scientific knowledge should be decided by crude, expertise-weighted, citation-weighted or Twitter-weighted vote counting.[28] Moreover, while both documents include a massive number of stellar scientists, the vast majority of the most influential scientists have not signed either document. Some of them may be embarrassed to sign given the adversarial, smearing environment that has emerged. Alternatively, many probably see that neither document contains the perfect truth. And, of course, many scientists generally abstain from signature collections.

Finally, while the data analysed here are limited to a relatively small number of top-cited scientists, the evaluation of the key scientific fields where these scientists publish offers some interesting hints. Both GBD and JSM include top signatories in disciplines such as epidemiology, public health and general and internal medicine

that are core pertinent fields in the pandemic. GBD has more diversity in field expertise and includes top signatories in quantitative disciplines such as statistics and bioinformatics, as well as paediatrics and ethics that are not represented among key JSM signatories, while JSM has superior representation in virology. These patterns may be due to chance given the relatively small sample analysed, and given the many thousands of additional signatories, these fields may well be represented in the longer lists. However, these patterns could also reflect some genuine differences in overall perspective between the two strategies. For example, GBD focuses more on the potential multifaceted collateral damage of lockdowns and on prioritising quantitative assessment of risk (where children and young people have far lower risk than elderly, vulnerable people),[29] while JSM depends more heavily on basic virology expertise. Given the magnitude of the COVID-19 crisis, it is important to ensure that scientific disciplines can collaborate dispassionately and that different views can be juxtaposed and integrated. GBD and JSM may have more in common than it is often thought. Critical differences between them should be probed with rigorous science rather than defended on partisan grounds and with social media warfare.

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