

# The spreading of misinformation online

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Edited by Matjaz Perc, University of Maribor, Maribor, Slovenia, and accepted by the Editorial Board December 4, 2015 (received for review September 1, 2015)

**The wide availability of user-provided content in online social media facilitates the aggregation of people around common interests, worldviews, and narratives. However, the World Wide Web (WWW) also allows for the rapid dissemination of unsubstantiated rumors and conspiracy theories that often elicit rapid, large, but naive social responses such as the recent case of Jade Helm 15—where a simple military exercise turned out to be perceived as the beginning of a new civil war in the United States. In this work, we address the determinants governing misinformation spreading through a thorough quantitative analysis. In particular, we focus on how Facebook users consume information related to two distinct narratives: scientific and conspiracy news. We find that, although consumers of scientific and conspiracy stories present similar consumption patterns with respect to content, cascade dynamics differ. Selective exposure to content is the primary driver of content diffusion and generates the formation of homogeneous clusters, i.e., “echo chambers.” Indeed, homogeneity appears to be the primary driver for the diffusion of contents and each echo chamber has its own cascade dynamics. Finally, we introduce a data-driven percolation model mimicking rumor spreading and we show that homogeneity and polarization are the main determinants for predicting cascades’ size.**

misinformation | virality | Facebook | rumor spreading | cascades

**T**he massive diffusion of sociotechnical systems and micro-blogging platforms on the World Wide Web (WWW) creates a direct path from producers to consumers of content, i.e., allows disintermediation, and changes the way users become informed, debate, and form their opinions (1–5). This disintermediated environment can foster confusion about causation, and thus encourage speculation, rumors, and mistrust (6). In 2011 a blogger claimed that global warming was a fraud designed to diminish liberty and weaken democracy (7). Misinformation about the Ebola epidemic has caused confusion among healthcare workers (8). Jade Helm 15, a simple military exercise, was perceived on the Internet as the beginning of a new civil war in the United States (9).

Recent works (10–12) have shown that increasing the exposure of users to unsubstantiated rumors increases their tendency to be credulous.

According to ref. 13, beliefs formation and revision is influenced by the way communities attempt to make sense of events or facts. Such a phenomenon is particularly evident on the WWW where users, embedded in homogeneous clusters (14–16), process information through a shared system of meaning (10, 11, 17, 18) and trigger collective framing of narratives that are often biased toward self-confirmation.

In this work, through a thorough quantitative analysis on a massive dataset, we study the determinants behind misinformation diffusion. In particular, we analyze the cascade dynamics of Facebook users when the content is related to very distinct narratives: conspiracy theories and scientific information. On the one hand, conspiracy theories simplify causation, reduce the complexity of reality, and are formulated in a way that is able to tolerate a certain level of uncertainty (19–21). On the other hand, scientific information disseminates scientific advances and exhibits the process of scientific thinking. Notice that we do not focus on the quality of the information but rather on the possibility of verification. Indeed,

the main difference between the two is content verifiability. The generators of scientific information and their data, methods, and outcomes are readily identifiable and available. The origins of conspiracy theories are often unknown and their content is strongly disengaged from mainstream society and sharply divergent from recommended practices (22), e.g., the belief that vaccines cause autism.

Massive digital misinformation is becoming pervasive in online social media to the extent that it has been listed by the World Economic Forum (WEF) as one of the main threats to our society (23). To counteract this trend, algorithmic-driven solutions have been proposed (24–29), e.g., Google (30) is developing a trustworthiness score to rank the results of queries. Similarly, Facebook has proposed a community-driven approach where users can flag false content to correct the newsfeed algorithm. This issue is controversial, however, because it raises fears that the free circulation of content may be threatened and that the proposed algorithms may not be accurate or effective (10, 11, 31). Often conspiracists will denounce attempts to debunk false information as acts of misinformation.

Whether a claim (either substantiated or not) is accepted by an individual is strongly influenced by social norms and by the claim’s coherence with the individual’s belief system—i.e., confirmation bias (32, 33). Many mechanisms animate the flow of false information that generates false beliefs in an individual, which, once adopted, are rarely corrected (34–37).

In this work we provide important insights toward the understanding of cascade dynamics in online social media and in particular about misinformation spreading.

We show that content-selective exposure is the primary driver of content diffusion and generates the formation of homogeneous

## Significance

**The wide availability of user-provided content in online social media facilitates the aggregation of people around common interests, worldviews, and narratives. However, the World Wide Web is a fruitful environment for the massive diffusion of unverified rumors. In this work, using a massive quantitative analysis of Facebook, we show that information related to distinct narratives—conspiracy theories and scientific news—generates homogeneous and polarized communities (i.e., echo chambers) having similar information consumption patterns. Then, we derive a data-driven percolation model of rumor spreading that demonstrates that homogeneity and polarization are the main determinants for predicting cascades’ size.**

Author contributions: M.D.V., A.B., F.Z., A.S., G.C., H.E.S., and W.Q. designed research; M.D.V., A.B., F.Z., H.E.S., and W.Q. performed research; M.D.V., A.B., F.Z., F.P., and W.Q. contributed new reagents/analytic tools; M.D.V., A.B., F.Z., A.S., G.C., H.E.S., and W.Q. analyzed data; and M.D.V., A.B., F.Z., A.S., G.C., H.E.S., and W.Q. wrote the paper.

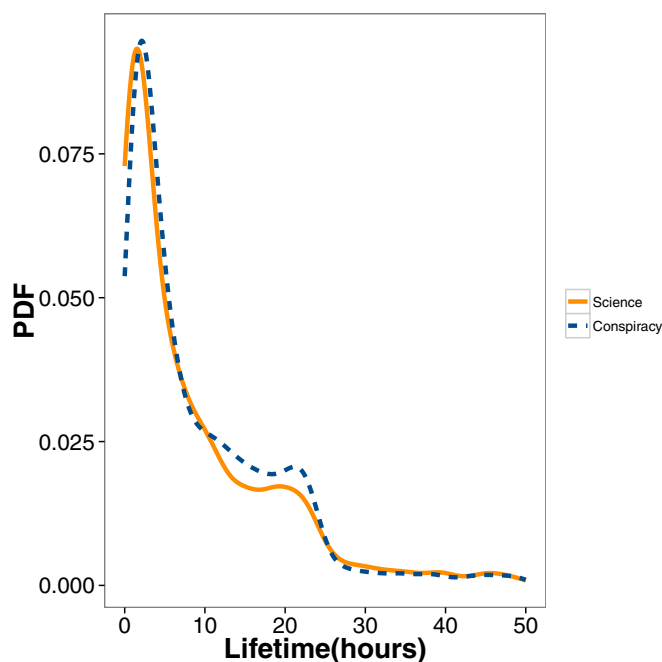
The authors declare no conflict of interest.

This article is a PNAS Direct Submission. M.P. is a guest editor invited by the Editorial Board.

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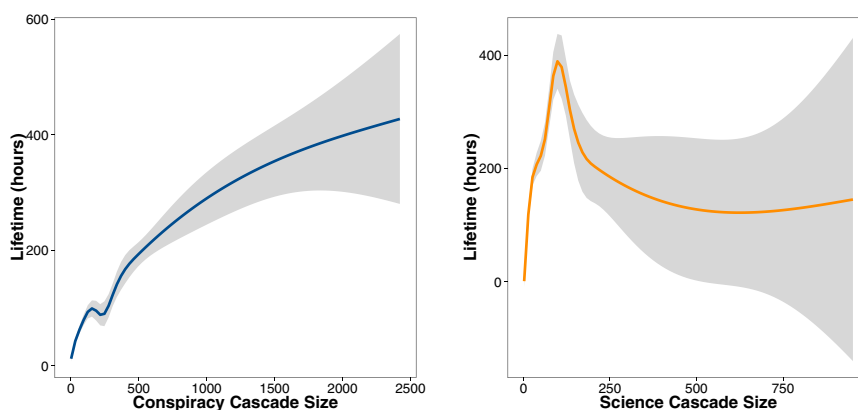
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This article contains supporting information online at [www.pnas.org/lookup/suppl/doi:10.1073/pnas.1517441113/-DCSupplemental](http://www.pnas.org/lookup/suppl/doi:10.1073/pnas.1517441113/-DCSupplemental).



clusters, i.e., “echo chambers” (10, 11, 38, 39). Indeed, our analysis reveals that two well-formed and highly segregated communities exist around conspiracy and scientific topics. We also find that although consumers of scientific information and conspiracy theories exhibit similar consumption patterns with respect to content, the cascade patterns of the two differ. Homogeneity appears to be the preferential driver for the diffusion of content, yet each echo chamber has its own cascade dynamics. To account for these features we provide an accurate data-driven percolation model of rumor spreading showing that homogeneity and polarization are the main determinants for predicting cascade size.

The paper is structured as follows. First we provide the preliminary definitions and details concerning data collection. We then provide a comparative analysis and characterize the statistical signatures of cascades of the different kinds of content. Finally, we introduce a data-driven model that replicates the analyzed cascade dynamics.



homogeneous if its edge homogeneity is positive. We then define a sharing path to be any path from the root to one of the leaves of the sharing tree. A homogeneous path is a sharing path for which the edge homogeneity of each edge is positive, i.e., a sharing path composed only of homogeneous links.

## Results and Discussion

**Anatomy of Cascades.** We begin our analysis by characterizing the statistical signature of cascades as they relate to information type. We analyze the three types—science news, conspiracy rumors, and trolling—and find that size and maximum degree are power-law distributed for all three categories. The maximum cascade size values are 952 for science news, 2,422 for conspiracy news, and 3,945 for trolling, and the estimated exponents  $\gamma$  for the power-law distributions are 2.21 for science news, 2.47 for conspiracy, and 2.44 for trolling posts. Tree height values range from 1 to 5, with a maximum height of 5 for science news and conspiracy theories and a maximum height of 4 for trolling. The resulting network is very dense. Notice that such a feature weakens the role of hubs in rumor-spreading dynamics. For further information see *SI Appendix*, section 2.1.

Fig. 1 shows the probability density function (PDF) of the cascade lifetime (using hours as time units) for science and conspiracy. We compute the lifetime as the length of time between the first user and the last user sharing a post. In both categories we find a first peak at  $\sim 2$  h and a second at  $\sim 20$  h, indicating that the temporal sharing patterns are similar irrespective of the difference in topic. We also find that a significant percentage of the information diffuses rapidly (24.42% of the science news and 20.76% of the conspiracy rumors diffuse in less than 2 h, and 39.45% of science news and 40.78% of conspiracy theories in less than 5 h). Only 26.82% of the diffusion of science news and 17.79% of conspiracy lasts more than 1 d.

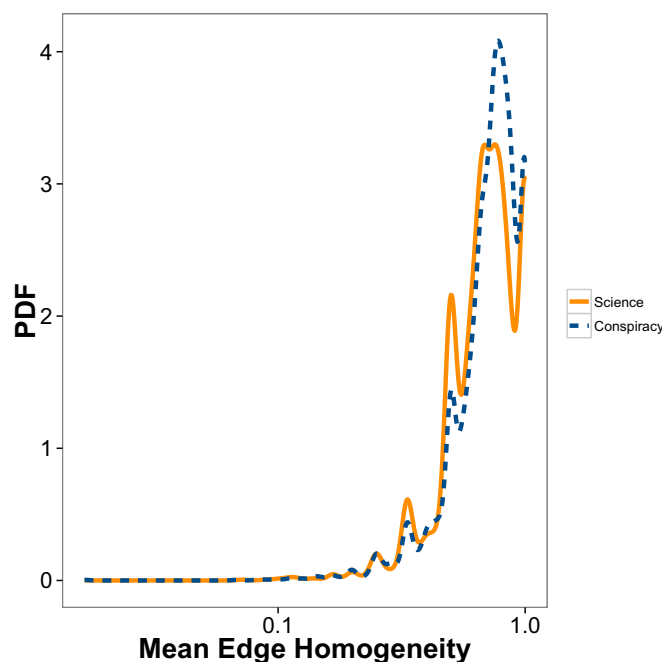
In Fig. 2 we show the lifetime as a function of the cascade size. For science news we have a peak in the lifetime corresponding to a cascade size value of  $\approx 200$ , and higher cascade size values correspond to high lifetime variability. For conspiracy-related content the lifetime increases with cascade size.

These results suggest that news assimilation differs according to the categories. Science news is usually assimilated, i.e., it reaches a higher level of diffusion quickly, and a longer lifetime does not correspond to a higher level of interest. Conversely, conspiracy rumors are assimilated more slowly and show a positive relation between lifetime and size. For both science and conspiracy news, we compute the size as a function of the lifetime and confirm that differentiation in the sharing patterns is content-driven, and that for conspiracy there is a positive relation between size and lifetime (see *SI Appendix*, section 2.1 for further details).

**Homogeneous Clusters.** We next examine the social determinants that drive sharing patterns and we focus on the role of homogeneity in friendship networks.

Fig. 3 shows the PDF of the mean-edge homogeneity, computed for all cascades of science news and conspiracy theories. It shows that the majority of links between consecutively sharing users is homogeneous. In particular, the average edge homogeneity value of the entire sharing cascade is always greater than or equal to zero, indicating that either the information transmission occurs inside homogeneous clusters in which all links are homogeneous or it occurs inside mixed neighborhoods in which the balance between homogeneous and nonhomogeneous links is favorable toward the former ones. However, the probability of close to zero mean-edge homogeneity is quite small. Contents tend to circulate only inside the echo chamber.

Hence, to further characterize the role of homogeneity in shaping sharing cascades, we compute cascade size as a function of mean-edge homogeneity for both science and conspiracy news (Fig. 4). In science news, higher levels of mean-edge homogeneity in



**Fig. 3.** PDF of edge homogeneity for science (orange) and conspiracy (blue) news. Homogeneity paths are dominant on the whole cascades for both scientific and conspiracy news.

the interval (0.5, 0.8) correspond to larger cascades, but in conspiracy theories lower levels of mean-edge homogeneity ( $\sim 0.25$ ) correspond to larger cascades. Notice that, although viral patterns related to distinct contents differ, homogeneity is clearly the driver of information diffusion. In other words, different contents generate different echo chambers, characterized by a high level of homogeneity inside them. The PDF of the edge homogeneity, computed for science and conspiracy news as well as the two taken together—both in the unconditional case and in the conditional case (in the event that the user that made the first share in the couple has a positive or negative polarization)—confirms the roughly null probability of a negative edge homogeneity (*SI Appendix*, section 2.1).

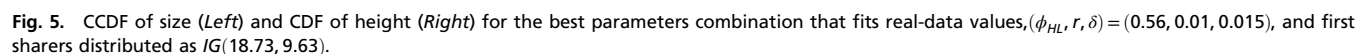
We record the complementary cumulative distribution function (CCDF) of the number of all sharing paths\* on each tree compared with the CCDF of the number of homogeneous paths for science and conspiracy news, and the two together. A Kolmogorov–Smirnov test and Q-Q plots confirm that for all three pairs of distributions considered there is no significant statistical difference (see *SI Appendix*, section 2.2 for more details). We confirm the pervasiveness of homogeneous paths.

Indeed, cascades' lifetimes of science and conspiracy news exhibit a probability peak in the first 2 h, and then in the following hours they rapidly decrease. Despite the similar consumption patterns, cascade lifetime expressed as a function of the cascade size differs greatly for the different content sets. However, homogeneity remains the main driver of cascades' propagation. The distributions of the number of total and homogeneous sharing paths are very similar for both content categories. Viral patterns related to contents belonging to different narratives differ, but homogeneity is the primary driver of content diffusion.

\*Recall that a sharing path is here defined as any path from the root to one of the leaves of the sharing tree. A homogeneous path is a sharing path for which the edge homogeneity of each edge is positive.







We find that the inverse Gaussian is the distribution that best fits the data both for science and conspiracy news, and for troll messages. For this reason, we performed one more simulation using the inverse Gaussian as distribution of the number of first sharers, 1,072 news items, 16,889 users, and the best parameters combination obtained in the simulations.<sup>8</sup> The CCDF of size and the CDF of height for the above parameters combination, as well as basic statistics considered, fit real data well.

Digital misinformation has become so pervasive in online social media that it has been listed by the WEF as one of the main threats to human society. Whether a news item, either substantiated or not, is accepted as true by a user may be strongly affected by social norms or by how much it coheres with the user's system of beliefs (32, 33). Many mechanisms cause false information to gain acceptance, which in turn generate false beliefs that, once adopted by an individual, are highly resistant to correction (34–37). In this work, using extensive quantitative analysis and data-driven modeling, we provide important insights toward the understanding of the mechanism behind rumor spreading. Our findings show that users mostly tend to select and share content related to a specific narrative and to ignore the rest. In particular, we show that social homogeneity is the primary driver of content diffusion, and one frequent result is the formation of homogeneous, polarized clusters. Most of the times the information is taken by a friend having the same profile (polarization)—i.e., belonging to the same echo chamber.

According to these settings algorithmic solutions do not seem to be the best options in breaking such a symmetry. Next envisioned steps of our research are to study efficient communication strategies accounting for social and cognitive determinants behind massive digital misinformation.

Values	Data	<i>I</i> G	Lognormal	Poisson
Min	1	0.36	0.10	20
First quantile	5	4.16	3.16	35
Median	10	10.45	6.99	39
Mean	39.34	39.28	13.04	39.24
Third quantile	27	31.59	14.85	43
Max	3,033	1814	486.10	66

<sup>§</sup>The best parameters combinations is  $\phi_{HL} = 0.56$ ,  $r = 0.01$ ,  $\delta = 0.015$ . In this case we have a mean size equal to 23.42 (33.43) and a mean height 1.28 (0.88), and it is indeed a good approximation: see [SI Appendix, section 3.2](#).

on the trash” for their valuable suggestions and discussions. Funding for this work was provided by the EU FET Project MULTIPLEX, 317532, SIMPOL, 610704, the FET Project DOLFINs 640772, SoBigData 654024, and CoeGSS 676547.

- PNAS | January 19, 2016 | vol. 113 | no. 3 | 559