

Opinion

Disinformation 2.0 in the Age of AI: A Cybersecurity Perspective

Why disinformation is a cyber threat.

ACCORDING TO A report from Lloyd's Register Foundation,^a at present, cybercrime is one of the biggest concerns of Internet users worldwide, with disinformation^b ranking highest among such risks (57% of Internet users across all parts of the world, socioeconomic groups, and all ages). For years, there has been a discussion in the security community about whether disinformation should be considered a cyber threat.¹⁰ However, recent worldwide phenomena (for example, an increase in the frequency and sophistication of cyberattacks, the 2016 U.S. election interference, the Russian invasion in Ukraine, the COVID-19 pandemic, and so forth) have made disinformation one of the most potent cybersecurity threats for businesses, governments, the media, and society as a whole. In addition, recent breakthroughs in AI have further enabled the creation of highly realistic fake content at scale. As such, we argue that disinformation should be rightfully considered a cyber threat, and therefore developing effective countermeasures is critically necessary.

The way disinformation is evolving

a The Lloyd's Register Foundation World Risk Poll (2019); <https://tinyurl.com/znjya6ct>

b In this column, among related concepts and terms, we use the term, *disinformation*, to refer to "false information created with malicious intention," per Kim et al.⁵



is not exactly new, as other "classical" cyber threats followed a similar path. First, disinformation has been around for centuries, and the Internet is just the latest means of communication used to spread it. We have already witnessed similar developments before. That is, there were different types of crimes such as scams, extortions, and thefts, and we now see their cyber versions, which are less risky for attackers yet more effective than their classical forms. Thus, the use of the Internet (especially social media) made it possible to boost the scale and range at

which these attacks can be launched, but the essence of the attack itself remains the same. Similarly, disinformation can affect many more people in a much shorter time than in the case of the non-digital version (for example, traditional newspapers, TV news). Moreover, advances in AI allowed the creation of deepfakes in various types of digital media (that is, images, video, speech) and text, and the introduced modifications are difficult to spot, distinguish, and explain. This greatly enhances the resulting disinformation's potential reach and believability.

Note that disinformation is not necessarily expected to provide direct revenue, as in the case of other cyber threats. However, such cases already have happened, for example, by spreading disinformation to manipulate stock price^c or earning income by disseminating it.^d

Disinformation 2.0

In the conventional paradigm of disinformation, on the attack side, we have disinformation creators who fabricate false information and post it to websites or social media for various purposes such as monetary incentives or political agenda. On the defense side, platforms have used human operators as well as computational methods to ensure the integrity of information, such as disinformation detectors to filter out questionable content, and socialbot detectors to curb the spread of disinformation. By and large, so far, creators (semi-) manually have created and disseminated disinformation content without using sophisticated AI techniques. When fake content is detected and filtered out by defense mechanisms, creators would simply attempt to redisseminate it using different accounts or domains.

With the recent advances in AI, we envision this paradigm is likely to change substantially, yielding what we call *disinformation 2.0*, where disinformation would become more targeted and personalized, its content indistinguishable from genuine content, and its creation and dissemination further accelerated by AI. Disinformation 2.0 will increase distrust in news that humans encounter in real and digital worlds, which is a major problem already.^e Typically, in cyberspace, an attacker's aim is to find weak spots in the targeted system and exploit/compromise them using technical and/or non-technical means. If we treat disinformation as a cyberattack, then several scenarios of disinformation 2.0 become possible, as illustrated in Figure 1.

1. Adversaries generate more convincing disinformation, using generative AI techniques (for example, ChatGPT for texts or DALL-E for images);

2. Adversaries subtly perturb existing content using AI methods to create more convincing disinformation with better capability to evade detection;⁶

3. Adversaries use AI methods to attack disinformation detection ecosystems, promoting disinformation and demoting real news;² and

4. Adversaries strategize the spread of disinformation using AI methods, maximizing its influence on the (social) network while evading socialbot detection ecosystems.⁷

When the creation and dissemination of disinformation are greatly enhanced using advanced AI methods in one of these scenarios, the resulting disinformation 2.0 becomes much more difficult to detect and more persuasive/impactful than the previous disinformation 1.0.

Countermeasures

Detecting disinformation 1.0 has been extensively researched in recent years (for example, see Kim et al.⁵ for a survey), and many solutions have reported high detection accuracies (for example, Cui et al.¹). However, there are still several remaining issues, including early detection, multilingual/multiplatform detection, better explainability, or sociotechnical issues.⁴ As with every cyber threat, completely eliminating disinformation is unlikely (as achieving complete security is never possible). We must diminish the impact of disinformation on Internet users, as we did with threats such as spam email messages. Several years ago, for instance, spam email messages were considered one of the major threats, but now their scale and relevancy are not as high as they were before.³ This has been achieved due to decades of research advances, during which many sophisticated techniques resulted in significantly limiting the volume of spam email messages in Internet users' inboxes.

Currently, a major disadvantage of our defenses against disinformation 2.0 is that they are being individually researched, developed, deployed, and evaluated, which is not very effective in diminishing the threat of disinformation.

c NBCnews. SEC Cracks Down on Fake Stock News. (2017); <https://nbcnews.to/4bi4uLj>

d H.C. Hughes and I. Waismel-Manor, The Macedonian Fake News Industry and the 2016 US Election; <https://bit.ly/3Os1G4B>

e Digital News Report 2022; <https://bit.ly/3UnJv3u>

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Figure 1. Four plausible attack scenarios under disinformation 2.0.

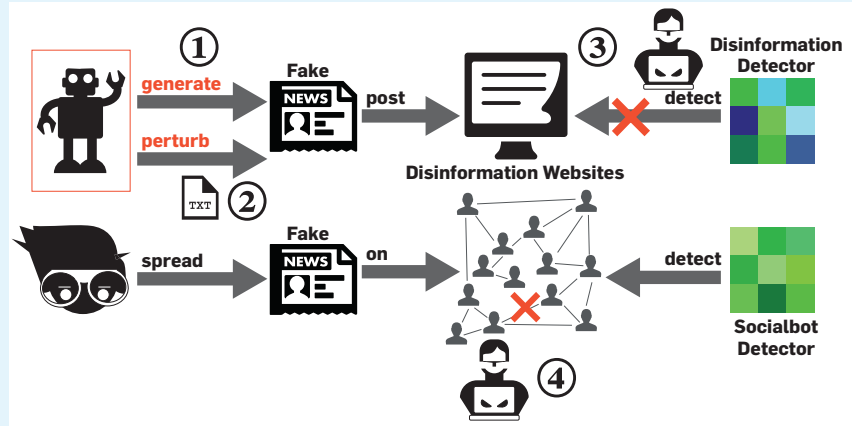
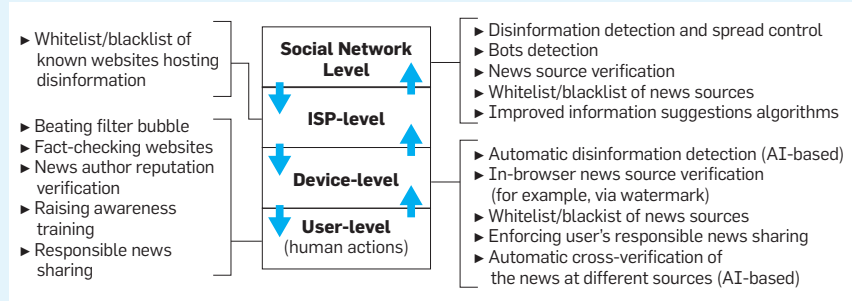


Figure 2. Proposed layered approach to counter disinformation in a holistic manner.



For this challenge, we argue to use some lessons learned from cybersecurity, where typically multiple “layers of defense” are envisioned. On the Internet, as security risks occur at various levels, it is necessary to set up security mechanisms that provide multiple layers of defense against these risks (for example, by considering threats at the system, network, application, and transmission levels).

A popular approach to layered security is defense-in-depth architectures, where controls are designed to protect the physical (that is, prevent physical access to IT systems, such as security guards or locked doors), technical (that is, security measures that use specialized hardware or software, for example, as a firewall, IDS/IPS, antivirus), and administrative (that is, policies or procedures directed at an organization’s employees) aspects of the communication network. Using such a layered approach to provide network security makes it possible for an attacker who penetrates one layer of defense to be stopped by a subsequent layer. Therefore, addressing

the problem is likely to require a layered approach, where both human-oriented (for example, raising awareness, training, and so forth) and technical measures are applied at the news creation, transmission, and consumption layers. We propose to distinguish four layers at which disinformation impact can be diminished (see Figure 2).

► *Social Network-level layer* is organized by the social network operator,

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which may be equipped with attack-resistant solutions like (AI-based and/or human operator-aided) disinformation detection, spread control mechanisms, and detection of strategized bots. Additional methods include, for example, verification of news sources (for instance, using reputation-based or digital watermarking-based solutions⁸), whitelisting and blacklisting of news sources, and fixing the information suggestion algorithms to avoid creating filter bubbles. This would be reminiscent of how network topology is taken into account in typical cybersecurity countermeasures for computer networks.

► *ISP-level layer* is organized at the Internet service provider (ISP), which is responsible for detecting, filtering, and blocking verified domains of disinformation 2.0 (this is already done, for example, for phishing email, suspicious links, or blacklisted domains). In such a scenario, the ISP can be considered a proxy between the users and the servers of the social network, located somewhere on the Internet.

► *Device-level layer* is organized on the user's machine, typically on a browser or mobile apps, as this is how the user interacts with various websites and social networks. The security mechanisms deployed on this level should include automatic (for example, AI-based) deepfake image or AI-generated text detection, in-browser news source verification and cross-verification of suspicious news at several trusted sources, and means to encourage responsible news sharing by the user (for example, alerting the user when he/she tries to spread the news marked as potentially fake).

► *User-level layer* is an essential part of a holistic approach to addressing disinformation 2.0, incorporating all manual actions that can be performed by users. For instance, this includes engaging with prebunking⁹ to raise the ability of the users to detect disinformation. Furthermore, given the rise and value of citizen journalism, it is important to empower users to perform disinformation detection using technical means that are commonly accessible to professional journalists in newsrooms. We consider the long-standing goal of educating the users about disinformation by empowering

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them is the best way to ensure their resilience in the long-term.

Note that to be effective, all security mechanisms and user actions must be applied in tandem. Moreover, employed mechanisms should be as diverse as possible, that is, they preferably should base their detection approaches on different aspects of disinformation. It is also worth emphasizing that currently, not all of the methods to fight disinformation described in this column are in use (for example, an automatic AI-based cross-verification of the news at different sources). Moreover, in Figure 2, arrows between the layers indicate that each layer can transfer certain information to the other layer. For instance, disinformation detection mechanisms on social networks can tag a video or image as questionable when passing the news to the user's browser/app for further probing by the next layer, or it can filter it out and send down only a proper notification. On the other hand, if disinformation is discovered on the user's device level, then this information can be passed to the social network operator and displayed to the user.

Some of the solutions in the proposed approach may be considered invasive from a user-privacy point of view, and protecting user data privacy is a critical concern in today's digital landscape. Fortunately, several effective solutions and strategies already exist that can be employed to fix this issue and follow the privacy-by-design principle, for example, by incorporating schemes relying on differential privacy, secure data aggregation, homomorphic encryption, or data masking and tokenization.

We strongly believe the advanced AI

techniques, despite their benefits to society, greatly enable adversaries to achieve more sophisticated and effective disinformation 2.0 attacks. As such, adopting the lessons learned from cybersecurity research, novel countermeasures are needed, especially a holistic layered approach as discussed. **C**

References

1. Cui, L. et al. DETERRENT: Knowledge guided graph attention network for detecting healthcare misinformation. In *Proceedings of the ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD)*. (2020), 492–502.
2. Du, Y. et al. Synthetic disinformation attacks on automated fact verification systems. In *Proceedings of the AAAI Conference on Artificial Intelligence (AAAI)*. (2022).
3. Ferrara, E. The history of digital spam. *Commun. ACM* 62, 8 (July 2019), 82–91.
4. Guo, Z. et al. A survey on automated fact-checking. *Transactions of the Association for Computational Linguistics* 10, 2, (2022); <https://bit.ly/3SFFBSI>
5. Kim, B. et al. A systematic review on fake news research through the lens of news creation and consumption: Research efforts, challenges, and future directions. *PLOS ONE* 16, 12 (Dec. 2021); 10.1371/journal.pone.0260080
6. Le, T. et al. Perturbations in the wild: Leveraging human-written text perturbations for realistic adversarial attack and defense. In *Findings of the Association for Computational Linguistics (ACL)*. (2022); 2953–2965.
7. Le, T. et al. Socialbots on fire: Modeling adversarial behaviors of socialbots via multi-agent hierarchical reinforcement learning. In *Proceedings of ACM Web Conference (WWW 2022)*; 545–554.
8. Megias, D. et al. Architecture of a fakenews detection system combining digital watermarking, signal processing, and machine learning. *J. Wireless Mobile Networks, Ubiquitous Computing, and Dependable Applications (JoWUA)* 13, 1 (Mar. 2022), 33–55; 10.22667/JOWUA.2022.03.31.033.
9. Roozenbeek, J. et al. Prebunking interventions based on the psychological theory of "inoculation" can reduce susceptibility to misinformation across cultures. *Harvard Kennedy School Misinformation Review* (Feb. 2020); 10.37016/mr-2020-008
10. Zurko, M.E. Disinformation and reflections from usable security. *IEEE Security and Privacy* 20, 3 (2022); 10.1109/MSEC.2022.3159405

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