

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.

^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>

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.

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Figure 1. Four plausible attack scenarios under disinformation 2.0.

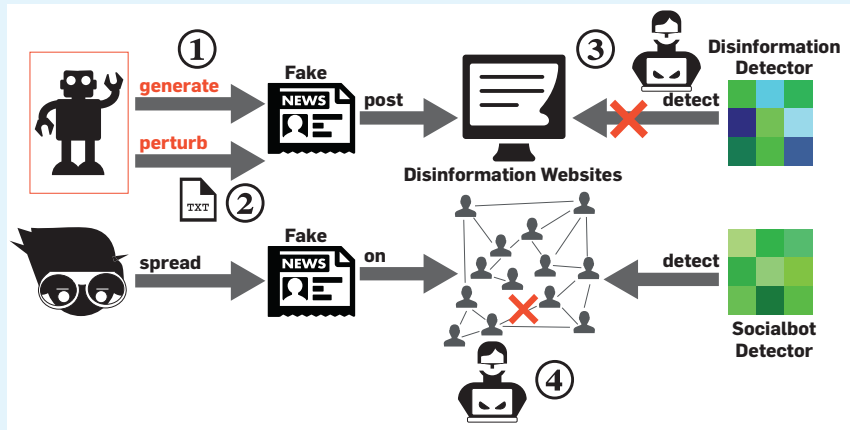
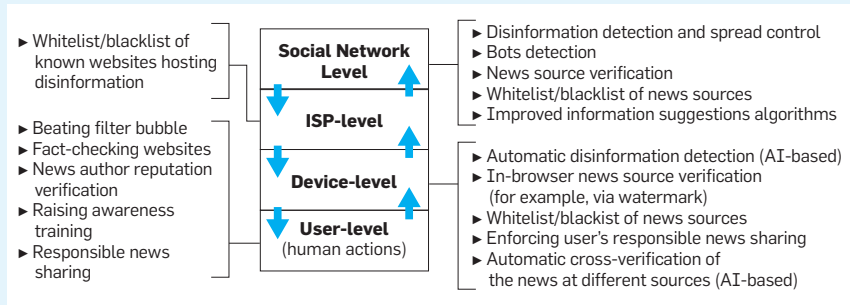


Figure 2. Proposed layered approach to counter disinformation in a holistic manner.



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,

The way disinformation is evolving is not exactly new, as other “classical” cyber threats followed a similar path.

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. **□**

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