

Comprehensive Review of Digital Harassment Prevention and Intervention Strategies: Bystanders, Automated Content Moderation, Legal Frameworks, AI, Education, Reporting, and Blocking

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Abstract

Digital harassment has become a significant issue in the digital age, enabled by the rise of online platforms that allow individuals to communicate freely and, often, anonymously. It encompasses various forms of harmful online behaviour, such as cyberbullying, trolling, doxing, and hate speech, which have been worsened by the increasing prevalence of social media and online communication platforms. Addressing this issue requires effective prevention strategies to ensure a safe online environment. This paper reviews strategies for cyberbullying prevention and intervention, including bystander intervention programs, automated content moderation, legislative measures, technological solutions like AI, educational programs and reporting mechanisms and blocking features. It analyses these strategies in terms of their focus areas, methods, key findings, challenges, and potential for future development, highlighting their implementation, outcomes, and areas for improvement based on existing literature.

Keywords: Digital Harassment, Bystander Intervention Programs, Automated Content Moderation, Legislative Approaches, Artificial Intelligence, Educational Programs, Reporting mechanisms, blocking features

1. Introduction

The digital age has made internet services accessible to people worldwide, from small children to senior citizens. The internet is widely used for various daily tasks, including sharing information, viewing online content, conducting financial transactions, running businesses through e-commerce platforms, making reservations, booking tickets, catering services, online learning and many more.

While the increasing integration of the internet into daily life has brought numerous benefits, it has also introduced new challenges, including digital harassment. Digital harassment involves the use of digital communication technologies to intimidate, harm, or coerce individuals. This harassment occurs on social media platforms, online games, websites, and through direct messaging. The problem is further aggravated by the anonymity provided by the internet and the rapid growth of social media.

Digital harassment forms includes Cyberbullying, Trolling, Revenge porn, Doxing, Online stalking. Cyberbullying refers to repeated acts of harassment, such as sending threatening messages, spreading



rumours, or deliberately excluding someone from online groups. Trolling means deliberately provoking others by posing inflammatory or offensive comments online. Revenge porn means the distribution of intimate images without the consent of the person involved. Doxing includes the public release of personal information such as email addresses, addresses or contact numbers. Online stalking refers to the repeated, obsessive attention directed towards an individual, including monitoring their online presence and activities.

Victims of digital harassment often suffer emotional distress, including anxiety, depression, and, in severe cases, suicidal thoughts. On a societal level, digital harassment erodes trust in online platforms, limits online freedom, and normalizes harmful behaviors in virtual spaces.

This research paper aims to analyses existing strategies for preventing digital harassment, including bystander intervention programs, automated content moderation, legislative measures, technological solutions like AI, educational programs and reporting mechanisms and blocking features - in terms of their focus areas, methods, key findings, challenges, and potential for future development.

2. Related Research Work

Digital harassment has been a focal point of research across multiple disciplines, including psychology, computer science, law, and education. Existing studies have explored various aspects of this issue, ranging from the psychological impact on victims to technological and policy-driven solutions aimed at prevention and intervention. We are now exploring existing research on strategies for the prevention and intervention of cyberbullying. These strategies encompass bystander intervention programs, automated content moderation, legislative measures, and technological solutions like AI, educational programs and reporting mechanisms and blocking features.

2.1 Bystander Interventions: A significant body of research has highlighted the effectiveness of bystander interventions in mitigating the impact of digital harassment. When we want to prevent bullying, many people can take an active role to intervene – especially bystanders. A bystander to cyber bullying is anyone who witnesses bullying either in person or in digital forms like social media, websites, text messages, gaming, and apps. Studies suggest that bystanders, when empowered and educated, can play a transformative role in interrupting harassment and providing support to victims. Many researchers suggested that bystanders are essential to bullying prevention and intervention. Bennet et al. 2021 highlights barriers to bystander intervention and suggested the need for awareness programs as cyber bullying prevention strategy. Cohen et al. 2019 discusses the importance of bystander intervention and strategies to empower them. The challenges were resistance from bystanders and lack of awareness. Espalage et al. 2012 identifies that when bullying occurs, bystanders are present 80 percent of the time. The challenges were psychological barriers as well as lack of training.

2.2 Automated content moderation: In the realm of technology-driven solutions, researchers have focused on the role of automated content moderation systems. Advances in artificial intelligence, machine learning and natural language processing to detect offensive or harmful content on online platforms. These tools are capable of analysing large volumes of data but may struggle with detecting context and subtleties in human language. Social media platforms use automated content moderation systems powered by machine learning algorithms to detect and filter harmful content, including hate speech, abusive language, and threats specified by Gillespie, T. 2018. A considerable body of research has focused on evaluating the effectiveness of automated moderation systems in preventing digital harassment. Studies suggest that while these tools can effectively detect explicit and obvious forms of harassment, they still face significant



challenges. A study by Zeng et al. 2020 found that existing hate speech detection systems have a false positive rate of over 30%, which undermines their effectiveness. A study by Binns et al. 2019 highlights how AI models fail to understand sarcasm, humour, and cultural differences in language, which can result in biased or inaccurate moderation.

2.3 Legislative Approaches: The importance of legal frameworks and policies has also been extensively discussed in the literature. Researchers have analysed how laws and regulations vary across regions and their effectiveness in deterring digital harassment. Many countries have passed laws aimed at preventing online harassment and bullying. Lang et al. 2019 reviewed the effectiveness of such laws in the United States and Europe. The research paper analyses the effectiveness of cyberbullying laws across different countries. These laws also vary widely between jurisdictions, which complicates enforcement, particularly on global platforms. Harris 2020 points out that in countries with weaker digital harassment laws, perpetrators often face no legal consequences, while in others, victims may have access to legal recourse through restraining orders or criminal charges. They provide global overview of digital harassment laws and challenges in implementation. While some studies highlight the success of stringent policies in reducing harassment, others point out limitations in enforcement and jurisdictional challenges in crossborder cases. They still faces challenge with Legal loopholes and jurisdictional issues.

2.4 Artificial Intelligence: AI technologies, particularly in Machine Learning (ML) and Natural Language Processing (NLP), have emerged as promising tools for detecting harassment. Research in this area has focused on enhancing the precision and scalability of AI-based interventions. AI-powered systems have shown promise in combating digital harassment by not only detecting harmful content but also predicting potential harassment events. Noves et al. 2020 explores opportunities and risks of AI in preventing online harassment. The research analyses online behaviour patterns to predict and prevent harassment before it occurs. These predictive models rely on large datasets to learn how digital harassment unfolds and flag problematic content or users pre-emptively. However, the ethical concerns around surveillance and privacy are significant issues. Liu et al. 2020 discusses how AI can help predict and prevent online harassment and argue that AI systems may infringe on user privacy and are not fool proof, as they rely heavily on the quality and biases in the data used to train them. Davidson et al., 2017 suggested that AI-powered systems are highly effective at identifying specific patterns of harassment, such as hate speech, using NLP models. However, challenge remains open for bias in datasets and AI models has also been a concern, with potential risks of unfairly targeting marginalized groups. Saxe and Berlin, 2015 discussed that these systems are not perfect in detecting sarcasm or evolving language. However, concerns about ethical implications, privacy, and transparency remain critical areas for further investigation.

2.5 Educational Programs: Educational programs and public awareness campaigns have been examined as preventative measures. Studies underscore the need for integrating digital literacy and empathy training into school curricula and community programs to foster a safer online environment. Wide research has been conducted for preventing cyberbullying and digital harassment on school-based anti-bullying programs. These programs aim to reduce bullying behaviors, improve school climate, and foster social-emotional development. Farrington & Ttofi, 2009 uses whole-school approach for effective programs which engages all members of the school community, including students, teachers, parents, and administrators. The study show that consistent enforcement of anti-bullying policies and teacher training improves program outcomes. The KiVa Program (Salmivalli et al., 2011) emphasizes empowering bystanders to intervene and support victims, which reduces bullying incidents. Programs like Second Step integrate SEL to teach empathy, emotion regulation, and conflict resolution. Research shows that SEL



helps reduce aggression and improve peer relationships (Espelage et al., 2015). With the increasing prevalence of online bullying, programs now incorporate digital citizenship and cyberbullying modules. Nonetheless, gaps remain in evaluating the long-term impact of these initiatives on reducing harassment incidents. Hinduja and Patchin, 2015 highlight the need for programs to address the unique challenges of online harassment.

2.6 Reporting mechanisms and blocking features: Platforms such as Facebook, Twitter, and Instagram have implemented user-driven tools for reporting harassment and blocking offenders. These tools enable users to protect themselves from harmful interactions. However, underreporting by victims and lack of effective enforcement of reporting measures are persistent issues (Towner, J., & McKay, D. 2021). Zhou et al. 2019 highlighted that reporting tools are underused, with many victims unsure of how to use them effectively.

3. Methodology and Key Findings

Addressing digital harassment requires a comprehensive understanding of various prevention and intervention strategies, each targeting unique aspects of this pervasive issue. Key areas of focus include the critical role of bystanders, the application of automated content moderation, the enforcement of cyberbullying laws, advancements in AI-powered detection systems, the effectiveness of school-based anti-bullying programs, and the importance of reporting mechanisms.

Each of these domains contributes to mitigating the impact of harassment in digital spaces. Research highlights the transformative potential of empowered bystanders, the limitations and biases of automated moderation tools, the necessity of harmonized legal frameworks, the promise of AI-driven solutions, the long-term benefits of structured educational programs, and the value of accessible reporting systems. The tables below summarizes the focus areas, Algorithms/Methods used and key findings of each of the studied digital harassment prevention and intervention strategies.

| Reference | Focus Area | Algorithms /Methods | Key Findings |
|---------------|--------------------|------------------------|--|
| Bennet S. et | Bystander | Psychological | Barriers include fear of retaliation and |
| al. (2021) | intervention in | models, behaviour | lack of confidence among bystenders |
| | digital harassment | analysis | lack of confidence among bystanders. |
| Cohen, L., et | Role of bystanders | Social network | Inclusive peer training improves |
| al. (2019) | in cyberbullying | analysis | bystander intervention. |
| Espalage, D., | Meta-analysis of | Behavioural | Bystander programs improve |
| et al. (2012) | bullying programs | outcome studies | intervention behaviour in schools. |

Table 1: Bystanders Interventions

Table 2: Automated Content Moderation

| Reference | Focus Area | Algorithms /Methods | Key Findings |
|-----------|--------------------|------------------------|-----------------------------|
| Ciller T | Content moderation | Moderation tools | Algorithmic moderation is |
| (2018) | and platform | (human + | insufficient without policy |
| (2018) | governance | automated) | enforcement. |



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| Reference | Focus Area | Algorithms /Methods | Key Findings | |
|--|-------------------|------------------------|-------------------------------------|--|
| Matias, J., & | Automated | ALassisted | Human moderators essential for | |
| Dastin, J. | moderation | moderation | numan moderators essential for | |
| (2020) | challenges | moderation | nualiced content. | |
| Zong V of al | Hata anaaah | | Fairness issues in hate speech | |
| $Z \in I $ $Z \in I $ $Z = 1$ $Z $ | datastion | Machine learning | algorithms due to language/cultural | |
| (2020) | detection | | biases. | |
| Binns, R., et al. | Bias in automated | A L and NL D | Context and language nuances pose | |
| (2019) | moderation | AI allu INLE | significant challenges. | |

Table 3: Cyberbullying Laws and Regulations

| Reference | Focus Area | Key Findings | |
|----------------------------------|--------------------|--|--|
| Lange \mathbf{P} at al. (2010) | Cyberbullying lowe | Cyberbullying laws show varying | |
| Lange, K., et al. (2019) | Cyberbullying laws | effectiveness based on enforcement. | |
| Harris I (2020) | Digital harassment | Need for harmonized digital harassment | |
| 11a1115, J. (2020) | laws (global) | laws worldwide. | |

Table 4: AI-Powered Detection Systems

| Reference | Focus Area | Algorithms /Methods | Key Findings |
|---------------------|------------------------|------------------------|----------------------------|
| Liu, Y., et al. | Predicting online | Doon loorning | Neural networks, sentiment |
| (2020) | abuse | Deep learning | analysis |
| Davidson, T., et | Hote speech detection | NI D | Offensive language |
| al. (2017) | Hate speech detection | INLE | classifiers |
| Saxe, J., & Berlin, | Malware detection | Doop noural nativorka | Malware detection |
| R. (2015) | (related tech concept) | Deep neural networks | algorithms |

Table 5: School-Based Anti-Bullying Programs

| Reference | Focus Area | Key Findings |
|--|------------------------|--|
| Farrington, D., & Ttofi, School-based bullying | | Programs reduce bullying over time with act |
| M. (2009) | programs | participation. |
| Salmivalli, C., et al. | ViVa program | Significant reduction in bullying with peer- |
| (2011) | Ki va program | focused models. |
| Espelage, D., & Swearer | Bullying prevention in | Highlighted long-term benefits of structured |
| (2011) | schools | education. |
| Hinduja, S., & Patchin, | Cyberbullying | Promotes ethical online behaviour through |
| J. (2015) | prevention | education. |
| | | |



| - | | | | |
|----------------------|-------------------------|---|--|--|
| Reference Focus Area | | Key Findings | | |
| Towner, J., & | Social media harassment | Enhanced reporting effectiveness with intuitive | | |
| McKay, D. (2021) | reporting tools | design. | | |
| Zhou, J., et al. | Paparting tools | Reporting tools need to be accessible and | | |
| (2019) | Keporung tools | anonymous for effectiveness. | | |

 Table 6: Reporting Mechanisms and Blocking Features

4. Limitations and Future Directions

While this review provides a comprehensive analysis of existing strategies for preventing and addressing digital harassment, several limitations warrant consideration. The research highlights challenges such as the limited generalizability of findings across diverse cultural and linguistic contexts, the evolving nature of digital harassment behaviors, and the dependency on self-reported data in many studies, which may introduce biases. Furthermore, technological solutions like AI-powered detection systems and automated content moderation face significant limitations, including algorithmic biases, contextual misunderstandings, and ethical concerns regarding privacy and transparency.

The tables below outlines the challenges and future directions associated with each prevention strategy. These future directions offer valuable insights and potential research ideas for further exploration in this field.

| Reference | Limitations | Future Directions |
|--------------------|------------------------------|---|
| Bennet, S., et al. | Psychological barriers not | Develop comprehensive bystander education |
| (2021) | addressed systematically. | programs integrating tech tools. |
| Cohen, L., et al. | Limited analysis of cultural | Explore gamified training for bystanders in |
| (2019) | variations in behaviour. | diverse cultural settings. |
| Espalage D et | Variability in program | Standardize program evaluations and enhance |
| al. (2012) | effectiveness across | scalability for diverse regions |
| | demographics. | scalability for diverse regions. |

Table 7: Bystanders Interventions

| Reference | Limitations | Future Directions |
|-------------------|---------------------------------|---|
| Gillespie, T. | Algorithms struggle with | Combine user reports with AI-driven |
| (2018) | nuanced context. | moderation for better accuracy. |
| Matias, J., & | Scalability of human | Invest in explainable AI and scalable hybrid |
| Dastin, J. (2020) | involvement is challenging. | moderation systems. |
| Zeng, X., et al. | Limited accuracy in diverse | Develop fairness-aware algorithms and |
| (2020) | linguistic contexts. | multilingual datasets. |
| Binns, R., et al. | Lack of linguistic diversity in | Expand linguistic and cultural representation |
| (2019) | training datasets. | in training data. |

| Table 8: | Automated | Content | Moderation |
|-----------|---------------------|---------|------------|
| I abic U. | <i>i</i> i utomatcu | Content | mouchanon |



| Reference | Limitations | Future Directions | | |
|--------------------|-----------------------------|---|--|--|
| Lange, R., et al. | Implementation gaps limit | Develop global frameworks for enforcement | | |
| (2019) | real-world effectiveness. | consistency. | | |
| Harris I (2020) | National differences hinder | Encourage international cooperation for | | |
| 11a1115, J. (2020) | global enforcement. | unified policies. | | |

Table 9: Cyberbullying Laws and Regulations

| Table 10. M-1 owered Detection Systems | | | |
|--|-------------------------------|---|--|
| Reference | Limitations | Future Directions | |
| Noyes, J., et al. | Privacy concerns and ethical | Focus on privacy preserving AI technologies | |
| (2020) | challenges with data use. | rocus on privacy-preserving Ar technologies. | |
| Liu, Y., et al. | Struggles with sarcasm, | Improve contextual understanding using | |
| (2020) | satire, and implicit abuse. | advanced NLP models. | |
| Davidson, T., et | Misclassification of context- | Train models on nuanced datasets with diverse | |
| al. (2017) | dependent language. | examples. | |
| Saxe, J., & Berlin, | Not directly applicable to | Adapt neural networks for harassment | |
| R. (2015) | digital harassment. | detection challenges. | |

Table 10: AI-Powered Detection Systems

Table 11: School-Based Anti-Bullying Programs

| Reference | Limitations | Future Directions |
|--------------------|-------------------------------|--|
| Farrington, D., & | Requires significant resource | Implement scalable solutions with tech support |
| Ttofi, M. (2009) | investment. | for smaller schools. |
| Salmivalli, C., et | Needs adaptation for cultural | Customize the KiVa model for global |
| al. (2011) | relevance in other countries. | scalability. |
| Espelage, D., & | Lacks focus on digital | Integrate digital safety education with |
| Swearer (2011) | bullying elements. | traditional bullying prevention. |
| Hinduja, S., & | Limited focus on | Combine educational approaches with |
| Patchin, J. (2015) | enforcement tools. | reporting and AI moderation tools. |

Table 12: Reporting Mechanisms and Blocking Features

| Reference | Limitations | Future Directions |
|------------------|------------------------------|--|
| Towner, J., & | Limited adoption by smaller | Expand design principles to all platform |
| McKay, D. (2021) | platforms. | scales. |
| Zhou, J., et al. | Underutilization due to lack | Increase awareness campaigns for using |
| (2019) | of awareness. | reporting tools effectively. |

5. Conclusion

In conclusion, digital harassment is a critical issue in the digital age, fuelled by the widespread use of online platforms that enable harmful behaviors such as cyberbullying, trolling, doxing, and hate speech. Effective prevention and intervention strategies are essential for fostering a safe online environment. This paper has reviewed various approaches, including bystander intervention programs, content moderation, legislative measures, technological advancements like AI, educational initiatives and Reporting and



blocking features. By analysing these strategies in terms of their focus areas, methodologies, key findings, challenges, and potential for improvement, this review underscores the importance of a multifaceted approach. Future research and development in these areas will play a crucial role in addressing the evolving challenges of digital harassment and ensuring a more secure and inclusive online space.

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