

# The Content Quality of Crowdsourced Knowledge on Stack Overflow- A Systematic Mapping Study

Gheida Shahrour  
Computer Science and Mathematics  
Keele University  
Newcastle-under-Lyme, UK  
g.shahrour@keele.ac.uk

Ed de Quincey  
Computer Science and Mathematics  
Keele University  
Newcastle-under-Lyme, UK  
e.de.quincey@keele.ac.uk

Sangeeta Lal  
Computer Science and Mathematics  
Keele University  
Newcastle-under-Lyme, UK  
s.sangeeta@keele.ac.uk

**Abstract**—Community Question Answering (CQA) forums such as Stack Overflow (SO) are a form of crowdsourced knowledge for software engineers who seek solutions to development and programming challenges. While such a forum provides valuable support to engineers, it often contains low quality content that impacts users' experience and the longevity of new users. Past research shows that most of the low-quality content comes from violating general Netiquette Rules (NRs). In the past, several researchers have worked on analysing the content of SO and suggested approaches to increase its quality. However, to the best of our knowledge, there is no previous work that has reviewed the scale of scientific attention that is given to this cause and the recommendations that have been made. We have conducted a Systematic Mapping Study (SMS) using five relevant databases, reviewing 1,489 papers and selecting 18 that are relevant to help to address this gap. We have found that SO has attracted increasing research interest on reducing NRs violations to improve the quality of communication on SO. Interestingly, the majority of papers used manual qualitative and quantitative analysis approaches to investigate this area. We have found that further research is required to identify more violation features, generalisable sources of data and that the use of computational analysis approaches are still needed in this area.

**Index Terms**—Stack Overflow, Crowdsourcing, Content Quality, Netiquette Rules violations, Systematic Mapping Study

## I. INTRODUCTION

Previous research has shown that social media has become unwelcoming [1] and that violating Netiquette Rules (NRs) and community norms drives users away from active engagement on platforms such as Twitter [2]–[5], Facebook [6] and on Community Question Answering (CQA) forums such as Gitter [1], [7], GitHub, Slack [1] and the focus of this paper, Stack Overflow (SO) [1]. The concept of netiquette assumes that online communication has its own set of rules such as Shea's NRs [8] and community-specific norms [9]. Using pleasantries such as apologies, over politeness, and gratitude are considered SO-specific norms (SO-Norms) violations [9].

Permission to make digital or hard copies of part or all of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. Copyrights for third-party components of this work must be honored. For all other uses, contact the Owner/Author.

ASONAM'23, November 6-9, 2023, Kusadasi, Turkiye  
© 2023 Copyright is held by the owner/author(s).  
ACM ISBN 979-8-4007-0409-3/23/11.  
<http://doi.org/10.1145/3625007.3627729>

As contributor participation is key for these platforms, content quality concerns including NRs and SO-Norms violations are becoming of increased importance and are difficult to detect and manage.

### A. Community Question Answering in Stack Overflow

Crowdsourcing has become an effective way of solving problems with individuals turning to the internet for help and solutions. Within computer science and software engineering, platforms including TopCoder [10], Bountify [11] and SO are available for practitioners to solve their domain specific problems. However, SO tends to be the preferred choice for practitioners [12] with more than 100 million users and 23.8 million questions placed as of June 2023 [13].

In order to increase the content quality of SO, human moderators manually detect and delete posts contaminated with NRs and SO-Norms violations. As the human moderation process is laborious, SO has introduced a Heat Detection (HD) bot in 2017<sup>1</sup>. If a post violates the NRs and SO-Norms, the HD bot flags the post and notifies the moderators. Moderators in turn assess the post and if the flagging appears to be legitimate, they delete the post. However, discussions between the askers and answers are written in natural language sentences which may contain hedges, sarcasm and emojis [14] that HD bot may find difficult to spot. Therefore, writing high-quality posts free of NRs and SO-Norms violations is a responsibility mostly left to SO practitioners.

Research has shown that violating NRs within social media impacts users' engagement and longevity in these platforms [1]–[7]. There have therefore been many efforts to detect NR violations and improve the quality of crowd-sourced platforms, but there has been limited attention to the scale and overall evaluation of this research area. Therefore, the aim of our research is to study the literature that has investigated the violations of different NRs set by Shea [8] and SO-Norms [9] in SO. With this aim, we carried out a Systematic Mapping Study (SMS) following the Kitchenham method [15], finding 18 relevant research papers focused on the violation of one or more aspects of NRs and SO-Norms in SO. We believe that

<sup>1</sup><https://stackoverflow.com/questions/7001/heat-detector-analysing-comments-to-find-heat>

this SMS will help in providing an overview of existing tools and techniques used to detect NRs and SO-Norms on SO and hence, provide a base for researchers to advance this field.

### B. Research Questions

This study provides a summary of the level of academic interest related to the content quality concerns such as NRs and SO-Norms which SO has generated, the publication venues that are targeted, approaches used and types of contributions that are provided. The following are the three research questions that we answered in this study:

**RQ1:** What are the types of NRs and SO-Norms violations in SO platform that are identified or reported in the selected studies?

**RQ2:** What level of academic interest related to NRs and SO-Norms violations has the SO platform generated over time?

**RQ3:** What are the methods utilised to investigate, identify, moderate or remove the NRs and SO-Norms violation on the SO platform in the selected studies?

The remaining sections of this paper are organised as follows. In Section 2, we provide details of the methodology used in conducting this SMS. In Section 3, we discuss the results obtained in this research. In Section 4, we discuss the main findings in our research study and their implications. In section 5, we provide potential future directions. Finally, in section 6, we summarise our SMS.

## II. METHOD

This study follows an SMS process which “provides a structure of the type of research reports and results that have been published by categorizing them” [16], an overview of research conducted, and helps identify research gaps.

### A. Manual and Automatic Search

In order to identify and refine the relevant keywords and terms used in the SMS, we began by manually searching digital databases using the terms ‘Stack Overflow’ and ‘quality concerns’. We then read the top ten papers and found that there are many published on ‘buffer overflow’ rather than ‘Stack Overflow’; consequently, we added the term ‘Community Question Answering’ to eliminate papers related to ‘buffer overflow’. We also limited the search to 2009-2023 due to SO being created at the end of 2008 [1]. Based on the results obtained and research questions formulated, the terms ‘Stack Overflow’, ‘Community Question Answering’ and ‘quality’ were identified. We then used these terms to search the following databases: Google Scholar, ACM Digital Library, IEEEExplore, Web of Science and ScienceDirect with records (1010, 80, 1262, 374 and 56 respectively). After reading the full text, we selected the first five papers from each database as our test set. If the test set appears in the final results using our automatic search string, then the string found is accurate.

To conduct our automatic search, we constructed the search strings as suggested by Kitchenham [15]: (“**Stack Overflow**’ OR ‘**StackOverflow**’ AND “**Community Question Answering**’ OR ‘**Community-based Question Answering**’ OR

‘**Community based Question Answering**’ OR ‘**Question and Answer**’ AND “**quality**’ OR ‘**standard**’ OR ‘**perform\***’ OR ‘**measure**’ OR ‘**benchmark**”). We inspected the title of the 1489 records and found that the test set appears in our automatic search. We selected the following databases: Google Scholar, ACM Digital Library, IEEEExplore, Web of Science and ScienceDirect in our search. Our selection follows the selections of other SMSs or systematic literature reviews in the software engineering discipline [17] [18]. Therefore, We used the same databases we used in our manual search above.

### B. Data Extraction

All publications that could not be used to answer our research questions were eliminated during the review process as suggested by Kitchenham [15]. Table I contains the inclusion and exclusion criteria which we applied to make these decisions more objective.

TABLE I  
INCLUSION AND EXCLUSION CRITERIA

Inclusion:	
In1	Is written in English
In2	Is published in peer reviewed venue
In3	Is fully accessible
In4	Is relevant to the research question in this SMS
In5	Is a conference or journal paper
Exclusion:	
Ex1	Appears repeatedly: only the latest version is considered
Ex2	Is a secondary study: systematic literature review or SMS
Ex3	Is a technical report, summary of conference, workshop, editorial, or keynote

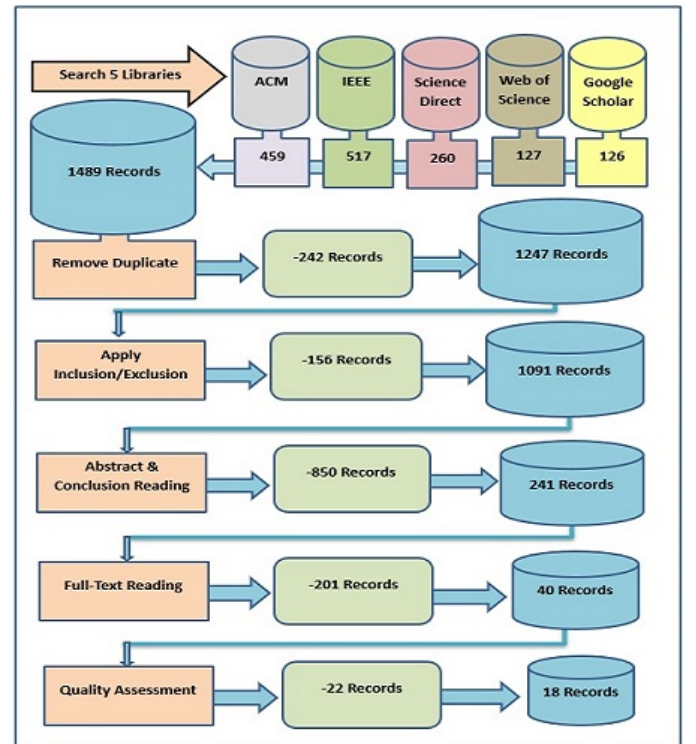


Fig. 1. Search Process and Total Papers Returned for Synthesis

From the 1489 results, we selected potentially relevant publications based on their meta data and full text. Once we removed the duplicates, we applied the inclusion and exclusion criteria to the remaining 1247 records, removing 156 records as shown in Figure 1. We then read the abstract and conclusion of the papers, removing 850 records that are not relevant to our research focus. We then read the full text of the remaining of 241 papers and found that only 40 papers were based on the content quality of SO. We then applied the quality test suggested by Kitchenham [15], conducted by the second and third authors of this paper and found a total of 18 publications ([1], [9], [19]–[34] respectively) that are relevant to our research focus.

### III. RESULTS

In the following subsections, we present our results according to our research questions described in Section B above.

#### A. Netiquette Rules and SO-Norms in Stack Overflow (RQ1)

The NRs and SO-Norms violations extracted from the selected papers are of two types based on the NRs practice be it negative and/or positive. As shown in Table II, there are 13 NRs and 6 SO-Norms violations found in the selected papers.

Results, in Table II, show that even in SO, a technical discussion forum, people may violate the generic NRs using abusive and offensive words very frequently, while words appertaining to racial abuse are not frequently discernible [9].

TABLE II  
TYPES OF NRs AND SO-NORMS VIOLATIONS

Taxonomy	Papers	Reference
<b>NRs Violations:</b>		
Subjectivity	P3	[20]
Moods, Harsh	P4	[21]
Offensive	P1 P4 P15	[1] [21] [9]
Neg. Emotions	P4 P6	[21] [23]
Gender Hostility	P5 P6 P11 P15	[22] [24] [28] [9]
Abusive Lang.	P9 P10 P15	[26] [27] [9]
Neg. Comments	P12 P13 P18	[29] [30] [34]
Hate Speech	P15	[9]
Rude Comments	P16	[32]
Human Values	P17	[33]
Hostility	P18	[34]
Norm Violations	P7	[24]
<b>SO-Norms Violations:</b>		
Gratitude	P2 P7 P8 P18	[19] [24] [25] [31]
Urgency, Reciprocity	P2	[19]
Pos. Emotions	P2 P3 P6	[19] [20] [23]
Politeness	P3 P6 P7 P8 P14	[20] [23] [24] [25] [31]
Pos. Comment	P12 P18	[29] [34]

Furthermore, users on SO accuse one another as being spam, sarcastic or rude, and moderators should take consequential action [9]. The moderation process, on the other hand, is found to be not instantaneous [9] as moderation requires multiple users flagging the content with NRs violations (e.g., harassment words) as offensive.

Although features under 'SO-Norms Violations' in Table II are considered positive or neutral in normal circumstances, pleasantries including apologies, welcome, over politeness,

and gratitude are considered SO-Norms violations and must be flagged [9]. As a consequence, moderators delete pleasantries as 'no longer needed'. This indicates that in technical forums such as SO, the majority of users do not like formality and over politeness. This is not the case with female users who prefer using pleasantries in their comments [31] in order to encourage answerers to generate higher quality content in the future, and increase users' engagements in SO [29].

#### B. Academic Interest: NRs and SO-Norms Violations (RQ2)

According to the 18 selected papers, no publication related to NRs and/or SO-specific norm violations in SO from 2009-2013 have been found. The research interest on NRs and SO-

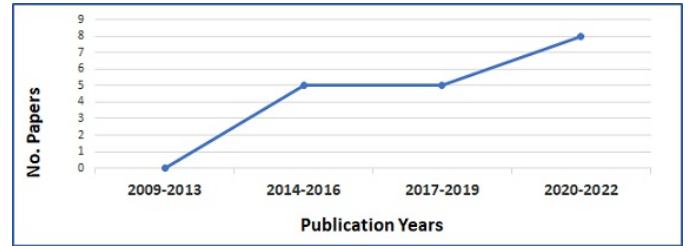


Fig. 2. Academic Interest related to NRs and SO-Norms Violations

Norms violations started in 2014 and gradually increased with most attention gained within the past three years. The research interest slowed down in 2017 when the 'Heat Detection' bot was embedded in SO in 2017 [1]. As shown in Figure 2, interest in this topic has increased in general though since 2014, and the evidence suggests that detecting and removing NRs SO-Norms violations in SO is currently an active topic.

#### C. Approaches Used in the Selected Papers (RQ3)

To answer RQ3 and for the scope of this paper, we looked into the approaches in general that are used in analysing NRs violations and found that 7 papers used manual qualitative and quantitative analysis methods which made a contribution of 38.89% in our SMS (depicted in row one of Table III).

TABLE III  
APPROACHES USED IN THE PAPERS

Papers	Approaches
P1	Random Forest and SVM
P7 P10 P13	Neural Network
P3 P8 P9	Regression Model
P11 P2 P4 P6	Machine Learning Tools
P12 P14 P15 P16 P17 P18	Quantitative/ qualitative Analysis
P5	Interviews

The next 61.11% of the papers are based on automated machine learning approaches such as Neural Networks [24], [27], [30], Logistic Regression Analysis [20], [25], [26], Random Forest and Support Vector Machine [1] and analysis tools such as SentiStrength [19], [27], [23]. While in [28], a machine learning technique tool (Scikit-learn) with clustering techniques are proposed to be used as future work.

#### IV. DISCUSSION

The aim of this SMS was to collect, analyse and interpret existing evidence pertaining to NRs and SO-Norms violations in SO. To the best of our knowledge, there is no previous systematic review or mapping study performed in this area. Out of 1489 papers found on content quality of SO, only 1.21% of papers were found to be related to NRs and SO-Norms violations in SO. Many interesting results are found in the selected papers representing NRs and SO-Norms taxonomy as shown in Table II and approaches to improve the content quality and consequently users' experience and engagement in SO as shown in Table III.

Many researchers investigated what and how community norms such as SO-Norms and NRs violations affect the content quality of SO [1], [9], [19]–[34]. Analysing the literature in this SMS, we found that there are two types of violations as shown in Table II: generic NRs violations and SO-Norms. Nine potential reasons that affect engagement of new users in SO [32] are mainly related to NRs violations. Results in our SMS show that users violate the generic NRs using unwelcoming or abusive words very frequently and racial abusive words less frequently [1]. Rude comments, for example, are flagged and deleted quickly by the moderators, but even in that situation, users end up reading these comments against them. This makes new users, who are not yet accustomed to the culture of SO, feel frustrated and unwelcomed, and consequently leading them to leave the community [32]. On the other hand, deleting low quality content can lead to negative behaviour, debates and frustrations explicit in users' comments [30]. Eventually, the negative behaviour could easily result in hurting SO's users, and apparently making the SO community hostile and unsupportive to new users. Therefore, there is a need of a prevention strategy to detect and delete NRs violations content from the users' comments [30].

In addition, positive comments and affective lexicon including apologies, wishes, overly politeness, jokes, gratitude are considered SO-Norms violations and must be flagged and deleted by moderators as 'no longer needed' [29]. 50% of comments containing gratitude expressions belong to new users of SO due to their unfamiliarity with SO's norm [25]. Calefato, et. al. found that regardless of user reputation (e.g., up/down vote count), successful answerable questions are the ones that have a neutral emotional style and do not contain abusive features such as unnecessary use of uppercase characters [26]. On the other hand, Jiarpakdee, et. al. found that community-based features (e.g., up/down vote count, offensive vote count, spam vote count) and affective lexicon play an important role in the question quality identification and question answerability [23]. Therefore, questions that are clear, direct interrogative, neither overly polite nor rude are more answerable than questions with an overly polite tone [20]. This indicates that in technical forums such as SO, the majority of users, except for females [31], do not like formality, gratitude and/or over politeness.

Despite the availability of HD bot and community modera-

tors in SO [1], many offensive contents bypass the moderators and the bot by obfuscated offensive words, and consequently offensive language still exists (0.14%) and 0.3% of the total comments are abusive. Therefore, many techniques and tools have been developed to detect and remove NRs and SO-Norms violations in SO. Novielli, et. al., for example, suggested a tool where emotional interface designs are proposed based on emotional intelligence that can be embedded in SO platforms [19]. Similarly, Maftouni, et. al. proposed a 'Support' button embedded in the SO that minimises toxic behaviour [34]. The button allows new users to report rude comments, express their opinions without waiting for a moderator to approve their report. In addition, as females post more apologetic contents in SO than men [31], the 'Support' button could help embrace women as well as users with more collectivist attitudes. Likewise, a 'Conflict Reduction System' is proposed to identify offensive words and suggest changes to minimise offence comments [1].

#### V. FUTURE WORK

The results shown in Table II are a list of certain NRs and SO-Norms violation features. For example, affective states (e.g., emotions, moods, harsh comments, avoid offensive behaviour, negative emotions) that presented in [21] are based on textual cues as the potential factors of success for an answer. However, there might be communication features other than the textual cues, Calefato, et. al. found that are not explored in this study using a dataset from SO as well as other CQA platforms [21]. In order to identify NRs and SO-Norms violation features, researchers could use emoticons [35] and emojis [36] features in addition to the textual cues. Moreover, to obtain robust identification results, researchers may fuse results from both the textual cues identification system and the emoji/ emoticon identification system.

Castelle, for example, found 23 comments contain offensive language but do not receive an 'offensive' flag [27] from users, human moderators or HD bot in SO. This could be due to users in SO write comments with hedging or abbreviations. In addition, many comments are considered offensive in context but not offensive when standing alone and thus, neural network methods perform poorly [27]. This indicates that there is a research gap in the automatic detection of offensive words that are hedged within positive words of posts in SO.

Another important result found in our SMS showed that the moderation process in SO is not instantaneous: to delete a post with NRs and SO-Norms violations content, moderators required responses from multiple users flagging the post as offensive. Nine comments with personal harassment words were not removed from SO [9]. While users reacted to NRs and SO-Norms violations in less than 2 hours, 76.35% of posts containing NRs and SO-Norms violation do not get downvoted at all. On the other hand, 148 of the original posters reacted to the concerns of potential NRs and SO-Norms violations by editing their posts. The slow process of flagging and moderation in SO inspires Chen, et. al. to adopt a

collaborative editing mechanism and fix posts with SO-Norms violations prior to posting them on the forum [24].

## VI. SUMMARY

This study aimed to build a comprehensive view of what the literature has reported on the factors that affect the content quality of SO, users' engagement and inclusivity. With a thorough selection process, a total of 18 out of 1489 papers were selected to investigate NRs and SO-Norms violations in SO. Our SMS indicated that there was an increase in studies on NRs and SO-Norms violations in SO and their influential factors. In addition, large datasets are needed not only to validate the results found in the selected papers, but also to identify communication cues including textual, emoji and emoticon embedded in CQA platforms. Furthermore, automated approaches are needed to validate the results in papers that used manual approaches.

## REFERENCES

- [1] J. Cheriyan, B. T. R. Savarimuthu, and S. Cranefield, "Towards offensive language detection and reduction in four software engineering communities," pp. 254–259, 2021.
- [2] E. Poole, E. H. Giraud, and E. de Quincey, "Contesting #stopislam: The dynamics of a counter-narrative against right-wing populism," *Open Library of Humanities*, 2019.
- [3] E. Poole, E. H. Giraud, and E. d. Quincey, "Tactical interventions in online hate speech: The case of #stopislam," *New Media Society*, vol. 23, no. 6, pp. 1415–1442, 2021.
- [4] A. A. Marouf, J. G. Rokne, and R. Alhajj, "Detecting and understanding sentiment trends and emotion patterns of twitter users - a study on the demise of a bollywood celebrity," *Big Data Cogn. Comput.*, vol. 6, p. 129, 2022.
- [5] K. Sailunaz and R. Alhajj, "Emotion and sentiment analysis from twitter text," *J. Comput. Sci.*, vol. 36, 2019.
- [6] K. Raiyani, T. Gonçalves, P. Quaresma, and V. B. Nogueira, "Fully connected neural network with advance preprocessor to identify aggression over facebook and twitter," in *TRAC@COLING 2018*, 2018.
- [7] B. T. R. Savarimuthu, Z. Zareen, J. Cheriyan, M. Yasir, and M. Galster, "Barriers for social inclusion in online software engineering communities - a study of offensive language use in gitter projects," *Proceedings of the 27th International Conference on Evaluation and Assessment in Software Engineering*, 2023.
- [8] S. SHEA, "Netiquette, alblion book," *SF CA*, 1994.
- [9] J. Cheriyan, B. T. R. Savarimuthu, and S. Cranefield, "Norm violation in online communities—a study of stack overflow comments," in *Coordination, Organizations, Institutions, Norms, and Ethics for Governance of Multi-Agent Systems XIII: International Workshops COIN 2017 and COINE 2020, Sao Paulo, Brazil, May 8-9, 2017 and Virtual Event, May 9, 2020, Revised Selected Papers*. Springer, 2021, pp. 20–34.
- [10] TopCoder, "Topcoder," 2023, accessed on 7 10, 2023. [Online]. Available: <https://www.topcoder.com/>
- [11] Bountify, "Bountify," 2023, accessed on 7 10, 2023. [Online]. Available: <https://bountify.co.usitestat.com/>
- [12] S. Meldrum, S. A. Licorish, and B. T. R. Savarimuthu, "Crowdsourced knowledge on stack overflow: A systematic mapping study," in *Proceedings of the 21st International Conference on Evaluation and Assessment in Software Engineering*, 2017, pp. 180–185.
- [13] A. Questions, "Stackoverflow," 2023, accessed on 7 10, 2023. [Online]. Available: <https://stackoverflow.com/questions/>
- [14] S. A. Alanazi, "Toward identifying features for automatic gender detection: A corpus creation and analysis," *IEEE Access*, vol. 7, pp. 111 931–111 943, 2019.
- [15] B. Kitchenham and S. Charters, "Guidelines for performing systematic literature reviews in software engineering," 2007.
- [16] K. Petersen, S. Vakkalanka, and L. Kuzniarz, "Guidelines for conducting systematic mapping studies in software engineering: An update," *Information and software technology*, vol. 64, pp. 1–18, 2015.
- [17] S. Meldrum, S. A. Licorish, and B. T. R. Savarimuthu, "Exploring research interest in stack overflow—a systematic mapping study and quality evaluation," *arXiv preprint arXiv:2010.12282*, 2020.
- [18] K. Nybom, A. Ashraf, and I. Porres, "A systematic mapping study on api documentation generation approaches," in *2018 44th Euromicro Conference on Software Engineering and Advanced Applications (SEAA)*. IEEE, 2018, pp. 462–469.
- [19] N. Novielli, F. Calefato, and F. Lanubile, "Towards discovering the role of emotions in stack overflow," in *Proceedings of the 6th international workshop on social software engineering*, 2014, pp. 33–36.
- [20] A. Y. Chua and S. Banerjee, "Answers or no answers: Studying question answerability in stack overflow," *Journal of Information Science*, vol. 41, no. 5, pp. 720–731, 2015.
- [21] F. Calefato, F. Lanubile, M. C. Marasciulo, and N. Novielli, "Mining successful answers in stack overflow," in *2015 IEEE/ACM 12th Working Conference on Mining Software Repositories*. IEEE, 2015, pp. 430–433.
- [22] D. Ford, J. Smith, P. J. Guo, and C. Parnin, "Paradise unplugged: Identifying barriers for female participation on stack overflow," in *Proceedings of the 2016 24th ACM SIGSOFT International symposium on foundations of software engineering*, 2016, pp. 846–857.
- [23] J. Jiarapakdee, A. Ihara, and K.-i. Matsumoto, "Understanding question quality through affective aspect in q&a site," in *Proceedings of the 1st International Workshop on Emotion Awareness in Software Engineering*, 2016, pp. 12–17.
- [24] C. Chen, X. Chen, J. Sun, Z. Xing, and G. Li, "Data-driven proactive policy assurance of post quality in community q&a sites," *Proceedings of the ACM on human-computer interaction*, vol. 2, no. CSCW, pp. 1–22, 2018.
- [25] Y. Fangl, T. Lu, P. Zhang, H. Gu, and N. Gu, "Exploring the effect of politeness on user contribution in q&a sites: A case study of stack overflow," in *2018 IEEE 22nd International Conference on Computer Supported Cooperative Work in Design ((CSCWD))*. IEEE, 2018, pp. 713–718.
- [26] F. Calefato, F. Lanubile, and N. Novielli, "How to ask for technical help? evidence-based guidelines for writing questions on stack overflow," *Information and software technology*, vol. 94, pp. 186–207, 2018.
- [27] M. Castelle, "The linguistic ideologies of deep abusive language classification," in *Proceedings of the 2nd workshop on abusive language online (ALW2)*, 2018, pp. 160–170.
- [28] S. Brooke, "'condescending, rude, assholes': Framing gender and hostility on stack overflow," in *Proceedings of the Third Workshop on Abusive Language Online*, 2019, pp. 172–180.
- [29] S. Sengupta and C. Haythornthwaite, "Learning with comments: An analysis of comments and community on stack overflow," in *Proceedings of the 53rd Hawaii International Conference on System Sciences*, 2020, pp. 2898–2907.
- [30] L. Tóth, B. Nagy, T. Gyimóthy, and L. Vidács, "Why will my question be closed? nlp-based pre-submission predictions of question closing reasons on stack overflow," in *Proceedings of the ACM/IEEE 42nd International Conference on Software Engineering: New Ideas and Emerging Results*, 2020, pp. 45–48.
- [31] E. Zolduoarrati and S. A. Licorish, "On the value of encouraging gender tolerance and inclusiveness in software engineering communities," *Information and Software Technology*, vol. 139, p. 106667, 2021.
- [32] A. Al Jobair, S. Mohammad, Z. R. Maisha, M. J. I. Mostafa, and M. N. Haque, "An empirical study on neophytes of stack overflow: How welcoming the community is towards them," in *ENASE*, 2022, pp. 197–208.
- [33] S. Krishtul, M. Shahin, H. O. Obie, H. Khalajzadeh, F. Gai, A. Rezaei Nasab, and J. Grundy, "Human values violations in stack overflow: An exploratory study," in *Proceedings of the 26th International Conference on Evaluation and Assessment in Software Engineering*, 2022, pp. 70–79.
- [34] M. Maftouni, P. M. J. Dubois, and A. Bunt, "'thank you for being nice': Investigating perspectives towards social feedback on stack overflow," in *Graphics Interface 2022*, 2022.
- [35] S. Bahri, P. Bahri, and S. Lal, "A novel approach of sentiment classification using emoticons," *Procedia computer science*, vol. 132, pp. 669–678, 2018.
- [36] S. Lal, N. Aggrawal, A. Jain, A. Khan, V. Tiwari, and A. Kaur, "Sentemojis: Sentiment classification using emojis," in *Information Management and Machine Intelligence: Proceedings of ICIMMI 2019*. Springer, 2021, pp. 293–300.