

METHODOLOGY

To identify the research works on mining software repositories, we adopted the search and selection process that involved renowned digital libraries, i.e., IEEE Xplore, ACM, Springer, Google Scholar, and ScienceDirect. Such libraries are selected because they are reputable and publish a significant number of SE-related journals and conference proceedings. For example, ACM and IEEE publish proceedings of the Conference on Mining Software Repositories. To query these libraries, we used pre-defined keywords including “Refactoring”, “Design Flaws”, “Code Smells”, “Change History”, “Commit Messages”, “Software History”, and “Mining Software Repositories”. Each library was searched by these keywords individually and in different combinations to find research works that fall within the scope of our study. The search strings were composed to contain important keywords that appear in most of the papers of our interest. For example, we formulated the following query strings to search the libraries.

1. “Mining change history to detect code smells”
2. “Mining software repositories to drive refactoring”
3. “Mining, code smells, refactoring, change history”

The following table highlights the results per query for each digital library. However, it is worth noting that, ACM was returning a big number of results and therefore we had to select the top few papers that appeared more relevant to our study.

SNo	Query	Digital Library	Number of outcomes
1	Mining change history to detect code smells	IEEE Xplore	7
		ACM	21,161
		Springer	844
		ScienceDirect	169
2	Mining software repositories to drive refactoring	IEEE Xplore	6
		ACM	243,894
		Springer	882
		ScienceDirect	102
3	Mining, code smells, refactoring, change history	IEEE Xplore	4
		ACM	289,333
		Springer	404
		ScienceDirect	95

LIST OF PAPERS COMPOSING OUR DATASET

1. AlOmar, E. A., Mkaouer, M. W., & Ouni, A. (2021). Toward the automatic classification of Self-Affirmed Refactoring. *Journal of Systems and Software*, 171. <https://doi.org/10.1016/J.JSS.2020.110821>
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11. Krasniqi, R., & Cleland-Huang, J. (2020). Enhancing Source Code Refactoring Detection with Explanations from Commit Messages. *SANER 2020 - Proceedings of the 2020 IEEE 27th International Conference on Software Analysis, Evolution, and Reengineering*, 512–516. <https://doi.org/10.1109/SANER48275.2020.9054816>
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13. Luaphol, B., Polpinij, J., & Kaenampornpan, M. (2022). Mining Bug Report Repositories to Identify Significant Information for Software Bug Fixing. *Applied Science and Engineering Progress*, 15(3). <https://doi.org/10.14416/J.ASEP.2021.03.005>
14. Mani, S. (2014). Improving enterprise software maintenance efficiency through mining software repositories in an industry context. *36th International Conference on Software Engineering, ICSE Companion 2014 - Proceedings*, 706–709. <https://doi.org/10.1145/2591062.2591085>
15. Nyamawe, A. S. (2022). Mining commit messages to enhance software refactorings

- recommendation: A machine learning approach. *Machine Learning with Applications*, 9, 100316. <https://doi.org/10.1016/J.MLWA.2022.100316>
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