How do various plagiarism detection tools differ in their detection results when applied to java/spring coding exercises

Dennis Goßler

Dennis Wäckerle

November 12, 2022

Abstract

Contents 2

Contents

1	Intro	oductio	vn	3				
2	How	/ Does	Plagiarism Detection Work	3				
3	Use	Case a	and Software Experiment	3				
	3.1	Use Ca	ase	3				
	3.2	Softwa	are Experiment	3				
4	Crit	eria for	Evaluation	3				
5	Eval	luation	of the different Tools	3				
	5.1	MOSS		3				
	5.2	JPlag		3				
		5.2.1	JPlag's comparison algorithm	4				
		5.2.2	The results	5				
		5.2.3	Integration into an automated evaluation pipeline	5				
	5.3 Plaggie							
		5.3.1	The Algorithm	5				
		5.3.2	The results	5				
		5.3.3	Integration into an automated evaluation pipeline	6				
	5.4	AC2		6				
		5.4.1	AC2 comparison algorithm	6				
		5.4.2	The results	6				
		5.4.3	Integration into an automated evaluation pipeline	6				
6	Con	clusion		6				
Re	eferen	ices		7				

1 Introduction

2 How Does Plagiarism Detection Work

Fand den Abschnitt bei moss ganz nett könnte man ja mit reinpacken Dennis G.

Moss and other plagiarism detection tools are not perfect, so a human should go other the results, and it should be checked if the clams are valid. "In particular, it is a misuse of Moss to rely solely on the similarity scores. These scores are useful for judging the relative amount of matching between different pairs of programs and for more easily seeing which pairs of programs stick out with unusual amounts of matching. But the scores are certainly not a proof of plagiarism. Someone must still look at the code." (Aiken, 2021)

3 Use Case and Software Experiment

3.1 Use Case

3.2 Software Experiment

4 Criteria for Evaluation

5 Evaluation of the different Tools

	C	C++	C#	Java	Kotlin	Python	PHP	VB.net	Javascript	Is expandable?
MOSS	Yes	Yes	Yes	Yes	No	Yes	No	Yes	Yes	No
JPlag	No	Yes	Yes	Yes	Yes	Yes	No	No	No	Yes
Plaggie	No	No	No	Yes	No	No	No	No	No	Yes ¹
AC2	Yes	Yes	No	Yes	No	Yes	Yes	No	No	Yes

Table 1 [The native supported programming languages for each plagiarism detection algorithm]

5.1 MOSS

Moss clams to be one of the best cheating detection algorithms. "The algorithm behind moss is a significant improvement over other cheating detection algorithms (at least, over those known to us)." (Aiken, 2021)

Moss supports the following programming languages C, C++, Java, C#, Python, Visual Basic, Javascript, FORTRAN, ML, Haskell, Lisp, Scheme, Pascal, Modula2, Ada, Perl, TCL, Matlab, VHDL, Verilog, Spice, MIPS assembly, a8086 assembly, a8086 assembly and HCL2

5.2 JPlag

JPlag is a plagiarism detection tool, which was developed by in 2000 at the University of Karlsruhe to help with the detection of plagiarized coding exercises. At that the tool was only available as a WWW service and could analyze C, C++. Scheme and Java programs (Perchelt et al., 2000, p. 4). Currently, the tool is available as an open source application which can

¹Plaggie is open source, therefore a custom tokenizer for different languages can be implemented.

be run locally and can be used with a CLI or a Java API. JPlag also currently supports 12 programming languages including the original four languages and more modern languages such as Kotlin ("JPlag - Detecting Software Plagiarism", 2022, Supported Languages) while also allowing to add new languages (Sağlam, 2022).

5.2.1 JPlag's comparison algorithm

JPlag's algorithm is split into two parts the tokenizing and the comparison of the token strings. During the tokenization process all programs are parsed and converted into token strings. In the second phase all token strings are compared in pairs to determine their similarity. This comparison uses a specially optimized version of the Greedy String Tiling algorithm. This algorithm tries to cover one token string with substrings of the other string. The percentage of the covered token strings is the similarity between the two programs (Perchelt et al., 2000, p. 10).

Tokenization The tokenization process is the only language dependent process of the JPlag algorithm (Perchelt et al., 2000, p. 10). The extracted tokens represent syntactic elements of the language like statements or control structures (Sağlam, 2022, How are submissions represented? — Notion of Token). During the parsing process each file is parsed with the result being a set of abstract syntax trees(AST) for each submission. Each AST is traversed depth first with nodes representing grammatical units of the language. During the traversal when entering and exiting a node a token can be created that match the type of the node and will then be added to the current list of tokens. There are block type nodes which can represent classes, if expressions or other elements of the language which have a corresponding beginning and end. When creating tokens from those nodes each have a corresponding "BEGIN" and "END" token. The token list should always have a pair of matching "BEGIN" and "END" tokens (Sağlam, 2022, How does the transformation work?).

Comparing token strings JPlag's comparison algorithm is essentially just the greedy string tiling algorithm for the comparison of two strings, however it is differently optimized to improve its runtime(Perchelt et al., 2000, p. 5). The goal of the algorithm is to find a set of substrings which are not only the same but also satisfy these three rules:

- 1. "Any token of A may only be matched with exactly one token from B."
- 2. "Substrings are to be found independent of their position in the string."
- 3. "Long substring matches are preferred over short ones[...]"

(Perchelt et al., 2000, p. 11)

These rule have some consequences. The first rule doesn't allow the matching of code that have been duplicated while the second rule makes the reordering of the source code not viable. Furthermore, the third rule is introduced since short matches are more likely to be spurious (Perchelt et al., 2000, p. 11).

- Applying the third rule sequentially leads to a greedy algorithm consisting of two phases

1. Two strings are searched for the biggest contigous matches. 3 nested Loops. Outer loop iterates over tokens in string A. Second loop compares Token T to Tokens in String B. Inner loop, if identical Tokens, extends the match as far as possible. Collects Set of longest common substrings 2. Marks all matches of maximal length. All tokens are marked and can not be used for further matches. Satisfies rule 1 - Repeat until no further matches found (Perchelt et al., 2000, p. 11) - tile is a unique and permanent association of a substring from A with a matching substring from B(Wise, 1993, p. 3) - MinimumMatchLength is defined must be atleast 1, should probably be higher, since it is unlikely that such a match will be significant (Wise, 1993, GST)

-similarity considers 100% if the shorter string is completly covered (Perchelt et al., 2000, p. 13)(See maths formula)

5.2.2 The results

- -comparisons without base code -successfully discovered plagiarism for m0 91.49% and 90.51% match -Other comparisons around 40% -need base code
- -m4 less clear 84.77%-76.2% -need base code and another submission

5.2.3 Integration into an automated evaluation pipeline

- -Can be executed locally -Has a cli and an api -> allows automation -Visualization in external web application -> might pose a problem
- -known unsuccessful attacks Changing the comments Changing the indentation Method and variable name changes(Ahtiainen, 2006, Known successful attacks)
- -know successful attacks Moving inline code to separate methods and vice versa Inclusion of redundant program code Changing the order of if-else blocks and case-blocks(Ahtiainen, 2006, Known unsuccessful attacks)

5.3 Plaggie

-Was the only open source tool when release(Ahtianien et al., 2006), JPlag now also open source

5.3.1 The Algorithm

-Uses CUP as a Parser -Uses standard LALR(1) parser generation("CUP", n.d.) -GST with no special optimization attempts -Algorithm was extended to support exclusion of common code(Ahtiainen, 2006, 4. Algorithm used)

5.3.2 The results

-TODO still have to run it.

6 Conclusion 6

5.3.3 Integration into an automated evaluation pipeline

-only Java 1.5(Ahtiainen, 2006; Ahtianien et al., 2006) -no further development since 2006 -has a cli -> can be automated -results as html files

5.4 AC2

"AC was born in the Escuela Politécnica Superior of the Universidad Autónoma de Madrid to deter and detect source-code plagiarism in programming assignments." (Freire, 2022)

5.4.1 AC2 comparison algorithm

The output that AC2 generates is visualization base. So AC it will not provide a value like "percentage of copy" instead, "it will create graphical representations of the degree of similarity between student submissions within a group" (Freire, 2022)

5.4.2 The results

5.4.3 Integration into an automated evaluation pipeline

6 Conclusion

References 7

References

- Ahtiainen, A. (2006). Readme for plaggie.
- Ahtianien, A., Surakka, S., & Rahikainen, M. (2006). *Plaggie: Gnu-licensed source code pla-giarism detection engine for java exercises*.
- Aiken, A. (2021). A system for detecting software similarity. Retrieved November 12, 2022, from https://theory.stanford.edu/~aiken/moss/
- Cup. (n.d.). Technische Universität München. Retrieved November 7, 2022, from http://www2.cs.tum.edu/projects/cup/
- Freire, M. (2022). Ac. https://github.com/manuel-freire/ac2
- Jplag detecting software plagiarism. (2022). JPlag. Retrieved November 12, 2022, from https://github.com/jplag/JPlag
- Perchelt, L., Malphol, G., & Phlippsen, M. (2000, March 28). *Jplag: Finding plagiarisms among a set of programs*.
- Sağlam, T. (2022, September 16). *4. adding new languages*. JPlag. Retrieved October 30, 2022, from https://github.com/jplag/JPlag/wiki/4.-Adding-New-Languages
- Wise, M. J. (1993). String similarity via greedy string tiling and running karp-rabin matching. University of Sydney, Department of Computer Science.