D Plagiarism Detection In Programming Language Source Codes Using NLP Tree Kernel Methods

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INTRODUCTION

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- Academic dishonesty is a universal problem (our institute included)
- Methods for finding duplicate code among peers generally works with heuristics based work or other language model based methods, which has proven to be ineffective in current scenario e.g. MOSS, JPlag, Sherlock
- Advantage of a computer program, over normal text: Well defined formal grammar
- Source codes are easy to process with the help of Abstract Syntax Tree and other flow diagrams using compiler toolchains.

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LIBRARIES USED

• CLANG

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- Clang is a compiler front end for the C, C++, Objective-C and Objective-C++ programming languages. It uses LLVM as its back end and has been part of the LLVM release cycle since LLVM 2.6.
- NLTK (http://www.nltk.org/)
 - NLTK is a leading platform for building Python programs to work with human language data.
- TREE-SVM (https://github.com/sitfoxfly/tree-svm)
 - Tree-SVM is used for Sub Tree, Subset Tree & Partial Tree Kernel extraction.
- Scikit-learn (http://scikit-learn.org/stable/)
 - Scikit-learn is used to apply SVM.





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Natural Language Analyses with NLTK



DATASET USED

- A set of C source code files, each containing multiple files for each user provided by mentors.
- Used MOSS (Measure Of Software Similarity) to assign tags to the set of plagiarized codes.

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CONCEPTS USED

- Language Model
 - Statistical Language Modelling is to build a statistical language model that can estimate the distribution of natural language as accurate as possible.
 - In-lining has been done to catch plagiarism, even if a person makes separate function for various parts of the code.

• AST transversal

• An abstract syntax tree is a tree representation of the abstract syntactic structure of the source code. Each node of the tree denotes a construct occurring in the source code.

• KL Divergence

• The Kullback–Leibler divergence, also known as information divergence, relative entropy is a non-symmetric measure of the difference between two probability distributions *P* and Q.

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THE PLAGIARISM DETECTION ARCHITECTURE USING LANGUAGE MODEL



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<u>APPROACH</u>

Language Model

 N-grams (up to trigrams) keyword comparison for C source codes using KL divergence. For bigrams matching technique, we need to consider the possibility that in a plagiarised code, a function of original code may have been written after in-lining. So before extracting keywords, we need to replace all function references in the codes with respective function declarations.

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Abstract Syntax Tree Matching

- Abstract Syntax Tree It provides the details about the basic structure of a code, such as Function declaration and Identifiers
- Generated using 'clang'

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• Also helps in avoiding the attempt of deceiving system by changing variable or function name





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WHAT IS NLP TREE KERNEL?

- Think of Kernel as a transformation
- Convert to the basic components

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 More suitable for source codes as generation of trees is easier using compiler toolchains



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TYPES OF KERNEL USED

1. Subtree Kernel (ST Kernel)

• The kernel returns a weighted sum of the number of common substrings (proper subtrees)

- 2. Subset Tree Kernel (SST Kernel)
 - The subset tree kernel (SST) defines a similarity measure between trees which is proportional to the number of shared subset trees
- 3. Partial Tree Kernel (PT Kernel)
 - A convolutional tree kernel, obtained by relaxing the constraint over SSTs that grammatical rules can't be broken

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Subtree Kernel



Subset Tree Kernel



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As 3 structures (out of 5) are completely identical the similarity is equal to 3

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CLASSIFICATION USING SVM

 Feature Vector: The 6 dimensional feature vector used for SVM contains 3 Kernel Similarity measures (Subset Tree, Sub Tree, Partial Tree) and KL Divergence between (File1 and File2), (File2and GeneralModel), (File2 and GeneralModel) between each pairs of files.

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CONFUSION MATRIX

	True Prediction	False Prediction	Total
Actual True Data (Plagiarism)	True Positive = 20	False Negative = 3	23
Actual False Data (No Plagiarism)	False Positive = 13	True Negative = 117	130
Total	33	120	

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Precision and Recall

Precision = true positives/(true positives + false positives) = 20/(20+13) = 60%
Recall = true positives/(true positives + false negatives) = 20/(20+3) = 87%

Comparison against baseline

- Applying SVM on kl divergence gives 50% accuracy on 10 fold cross validation.
- Whereas SVM on kl divergence with Kernel Tree gives 78% accuracy on 10 fold cross validation.

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 The user needs to paste the source code files to check for plagiarism detection in a specified folder as shown in the figure.

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OUTPUT FORMAT

The pair of plagiarised codes will be shown in an output.txt file created in the project folder.



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<u>CITATIONS</u>

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