

Abstract

In this poster we present an automated method for empirically evaluating clone detection tools. Our method leverages mutation-based techniques to overcome existing limitations of tool evaluation studies by automatically synthesizing large numbers of known clones based on an editing theory of clone creation. Our framework is effective in measuring recall and precision of clone detection tools for various types of fine-grained clones in real systems without manual intervention.

1 Introduction

- Copying a code fragment and reusing it by pasting with or without minor modifications is a common practice in software development.
- Consequently, software systems often have duplicated code (between 7% and 23%) [2, 8].
- However, duplication is harmful for software maintenance and evaluation (e.g., bug propagation) [8].
- Thus, need to detect these “code clones”.
- Fortunately, a great many tools have been proposed [7, 8].

Motivation:

- This huge number of tools calls for quantitative evaluations and there have been several.
- But there is a lack of a validated clone benchmark.
- And huge manual effort is required to hand check large numbers of candidate clones.
- Bellon et al. [3] is the most extensive one to date, but only 2% of clones were oracled and many parameters may have influenced the results [2].
- Thus, we propose a controlled experiment, using the well known mutation analysis technique [1] from the testing community.

2 Background

Code Fragment: A code fragment (CF) is any sequence of code lines and is identified by its file name and begin-end line numbers, denoted as a triple $\langle CF.FileName, CF.BeginLine, CF.EndLine \rangle$.

Code Clone: A code fragment CF_2 is a clone of another code fragment CF_1 if they are similar by some given definition of similarity, that is, $f(CF_1) = CF_2$ where f is the similarity function (see clone types below).

Clone Pairs/Classes: Two fragments that are similar to each other form a clone pair (e.g., $\langle CF_1, CF_2 \rangle$), and when many fragments are similar, they form a clone class or clone group.

In this example, five clone pairs, $\langle F1(a), F2(a) \rangle$, $\langle F1(b), F2(b) \rangle$, $\langle F2(b), F3(a) \rangle$, $\langle F2(c), F3(b) \rangle$ and $\langle F1(b), F3(a) \rangle$. But actually three clone pairs, $\langle F1(a+b), F2(a+b) \rangle$, $\langle F2(b+c), F3(a+b) \rangle$ and $\langle F1(b), F3(a) \rangle$. One clone class is $\langle F1(b), F2(b), F3(a) \rangle$.

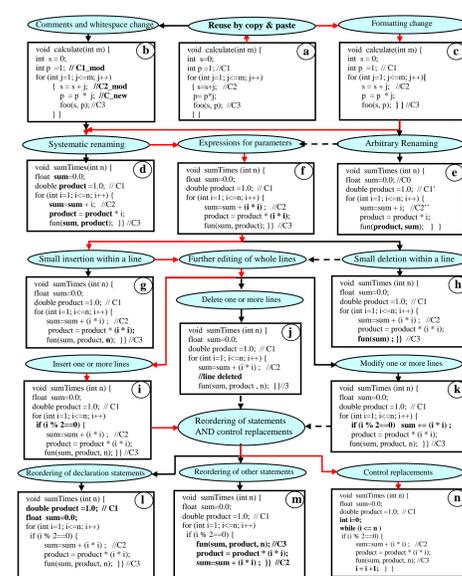
Fragment 1:	Fragment 2:	Fragment 3:
for (int i=1; i<n; i++) { sum = sum + i; }	for (int i=1; i<n; i++) { sum = sum + i; }	...
if (sum < 0) { sum = n - sum; }	if (sum < 0) { sum = n - sum; }	if (result < 0) { result = m - result; }
...	while (sum < n) { sum = n / sum; }	while (result < m) { result = m / result; }
...

Clone Types: The definition of clone is inherently vague in the literature [8]. However, the following four types can roughly be defined [3, 8].

- Type 1:** Identical code fragments except for variations in whitespace, layout and comments.
- Type 2:** Syntactically identical fragments except for variations in identifiers, literals, types, whitespace, layout and comments.
- Type 3:** Copied fragments with further modifications such as changed, added or removed statements, in addition to variations in identifiers, literals, types, whitespace, layout and comments.
- Type 4:** Two or more code fragments that perform the same computation but are implemented by different syntactic variants.

3 Editing Taxonomy and Mutation Operators for Cloning

- For any mutation-based analysis, availability of a set of representative mutation operators is a primary concern.
- For example, numerous mutation generators are available for generating potential “bugs” in various languages [1].
- However, mutation operators for code cloning have to our knowledge NOT been studied so far.
- Thus, we have designed an editing taxonomy of different clone types [7] by studying the literature [8]. The taxonomy is also validated by studying the copy/paste patterns of the function clones [5] in our empirical studies [10].
- This taxonomy has been used to design mutation operators for cloning.



In the following table we list the different mutation operators.

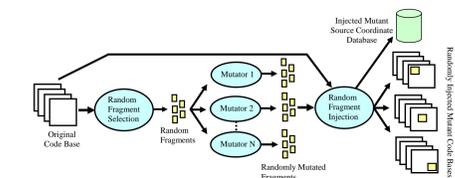
Name	Random Editing Activities	Ref**	Type
mCW	Changes in whitespace and	a => b	Type 1
mCC	Changes in comments	a => c	Type 1
mCF	Changes in formatting	a => c	Type 1
mSRI	Systematic renaming of identifiers*	c => d	Type 2
mARI	Arbitrary renaming of identifiers*	c => e	Type 2
mRPE	Replacement of identifiers with expressions	d => f	Type 2
mSIL	Small insertions within a line	f => g	Type 3
mSDL	Small deletions within a line	f => h	Type 3
mILs	Insertions of one or more lines	g => i	Type 3
mDLs	Deletions of one or more lines	g => j	Type 3
mMLs	Modifications of whole line(s)	g => k	Type 3
mRDS	Reordering of declarations	i => m	Type 4
mROS	Reordering of other statements+	i => l	Type 4
mCR	Replacing one type of control by another	i => n	Type 4

- Each mutation operator performs single level editing as of the table above.
- Combination of the mutation operators has been used to perform multiple level editing, e.g. (Original Fragment (a) => Formatting change (c) => systematic renaming of identifiers (d) => expressions for parameters (f) => small insertion within a line (g) => insertion of new lines (i) => control replacement (n) by following the solid (red) lines on the example taxonomy.
- The source transformation language TXL [4] is used to implement the mutation operators.

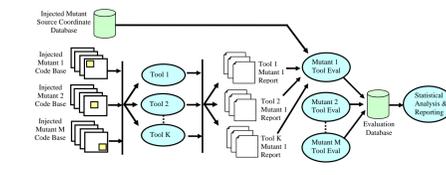
4 The Framework

The framework has two main phases as follows:

Generation Phase: Randomly mutated clone fragments are generated from the original code base and randomly injected into the code base to get mutated code bases.



Evaluation Phase: The mutated code bases are used to evaluate and compare clone detection tools. The random mutation and injection steps allow for thousands of randomly placed clones to be generated.



5 Measurement of Recall

Recall definition is the usual one in IR research, that is, the number of items detected divided by the total number of detectable items.

If the mutant clone $moCF$ of original code fragment oCF injected into mutant code base $mioCB$ of code base oCB is “killed” (i.e., $\langle oCF, moCF \rangle$ is detected as a clone pair) by the detector, then its recall for that clone is 1, otherwise it is 0. We can denote this decision by:

$$R_T^{(oCF, moCF)} = \begin{cases} 1, & \text{if } \langle oCF, moCF \rangle \text{ is detected by } T \text{ in } mioCB; \\ 0, & \text{otherwise.} \end{cases}$$

The same mutated code fragment, $moCF$ can be randomly injected to the original code base, oCB any number of times, producing n different mutated / injected versions of oCB , say $mioCB_1, mioCB_2 \dots mioCB_n$.

The random fragment selector chooses m code fragments (say $oCF_1, oCF_2 \dots oCF_m$) from the code base, and each of them will be mutated by each mutation operator $dmOP$ producing mutated code fragments $moCF_1, moCF_2 \dots moCF_m$.

Thus, the recall for mutation operator $dmOP$ for tool T is given by:

$$R_T^{dmOP} = \frac{\sum_{i=1}^{n*m} R_T^{(oCF_i, moCF_i)}}{m * n}$$

Similarly, recall of tool T for Type 1 clones (three mutation operators and their combinations) can be defined as:

$$R_T^{Typ1} = \frac{\sum_{i=1}^{n*m*(3+4)} R_T^{(oCF_i, moCF_i)}}{m * n * (3 + 4)}$$

The overall recall for tool T is the summary of recall for the l clone mutation operators and c combinations applied n times to m selected code fragments, given by:

$$R_T^{overall} = \frac{\sum_{i=1}^{n*m*(l+c)} R_T^{(oCF_i, moCF_i)}}{m * n * (l + c)}$$

6 Measurement of Precision

Using the notation of the previous subsection, let us say that for a mutated code fragment $moCF$ created by mutation operator $dmOP$, a tool T reports k clone pairs, $\langle moCF, CF_1 \rangle, \langle moCF, CF_2 \rangle \dots \langle moCF, CF_k \rangle$ in mutant code base $mioCB$.

If automatic validation reports that v of these are valid, then the unit precision of the tool T for the single injection of $moCF$ for clone type/mutation operator $dmOP$ is as follows:

$$P_T^{dmOP \text{ w.r.t. single injection of } moCF} = \frac{v}{k}$$

Similarly to the previous section, the precision for mutation operator $dmOP$, for Type 1 and for a tool T (overall precision) is given by:

$$P_T^{dmOP} = \frac{\sum_{i=1}^{n*m} v_i}{\sum_{i=1}^{n*m} k_i}$$

$$P_T^{Typ1} = \frac{\sum_{i=1}^{n*m*(3+4)} v_i}{\sum_{i=1}^{n*m*(3+4)} k_i}$$

$$P_T^{overall} = \frac{\sum_{i=1}^{n*m*(l+c)} v_i}{\sum_{i=1}^{n*m*(l+c)} k_i}$$

7 Mapping of Code Fragments

We say that a code fragment CF_1 is contained by another fragment CF_2 if both are in the same file, and the range of line numbers of CF_1 is within the range of line numbers of CF_2 . In algorithmic form,

```
boolean isContained(CF CF1, CF CF2) {
    return (CF1.FileName == CF2.FileName)
        AND (CF1.BeginLine >= CF2.BeginLine)
        AND (CF1.EndLine <= CF2.EndLine)
}
```

The following algorithm implements our definition of detection for a mutant pair (MP) consisting of clone mutant $moCF$ of original fragment oCF and a tool T 's clone candidate set (CSet) of detected clone pairs C :

```
boolean isDetected(MP (oCF, moCF), CSet C) {
    for each clone pair (CF1, CF2) in C {
        if ((isContained(oCF, CF1)
            AND isContained(moCF, CF2))
            OR (isContained(moCF, CF1)
            AND isContained(oCF, CF2)))
            return True;
    }
    return False;
}
```

To measure precision, we need to find all pairs in C for which one of the fragments is the mutant clone $moCF$:

```
CSet validateUs(CF moCF, CSet C) {
    CSet ValidateMe = {};
    for each clone pair (CF1, CF2) in C {
        if (isContained(moCF, CF1)
            OR isContained(moCF, CF2))
            ValidateMe = ValidateMe + (CF1, CF2);
    }
    return ValidateMe;
}
```

8 Validation of Clone Pairs

- Recall measurement is completely automatic and no validation of clone pairs is required.
- But, to accurately measure precision we need to validate those few clone pairs that are associated with the mutant code fragment.
- We develop a clone validator based on our NICAD [6, 10] tool.
- The validator is well aware of the mutation operators applied and changes made on the cloned fragment, and thus can accurately measure their real similarity.

9 Adapting Tools to the Framework

- The tool should be run from the command line (most tools usually do)
- It should provide a textual report of the detected clones, the usual column-oriented textual format of full file name and begin/end line numbers of the code fragments of the candidate clone pairs.

10 Conclusion

- Existing studies for empirically evaluating clone detection tools have had several limitations.
- We have designed a new approach for evaluating clone detection tools in a controlled way by borrowing an established technique from the testing community – mutation-based analysis.
- An experiment was successfully conducted with three different variants of our NICAD tool.
- The framework can run subject tools with varying tunable parameters suitable for identifying different types of clones.
- The framework is language-specific, and currently supports only C, Java and C#. However, since the majority of the framework is language-independent, it is not difficult to add new languages.
- Detailed description of the framework can be found at [11] and an earlier outline at [9].

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