EvoSuite at the SBST 2016 Tool Competition

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ABSTRACT

EVOSUITE is a search-based tool that automatically generates unit tests for Java code. This paper summarizes the results and experiences of EVOSUITE's participation at the fourth unit testing competition at SBST 2016, where EVO-SUITE achieved the highest overall score.

CCS Concepts

•Software and its engineering \rightarrow Software testing and debugging; Search-based software engineering;

Keywords

test case generation; search-based testing; testing classes; search-based software engineering

1. INTRODUCTION

This paper describes the results of applying the EVOSUITE test generation tool [2] to the benchmark used in the tool competition at the International Workshop on Search-Based Software Testing (SBST) 2016. Details about the competition and the benchmark can be found in [14]. In this competition, EVOSUITE achieved a 1126.7 overall score, which was the highest among the competing tools.

2. ABOUT EVOSUITE

EVOSUITE [2, 6] automatically generates test suites for Java classes, targeting branch coverage and other coverage criteria (e.g., mutation testing [8]). EVOSUITE works at the Java bytecode level, i.e., it does not require source code. It is fully automated and requires no manually written test drivers or parameterized unit tests. For example, when EVO-SUITE is used from its Eclipse and IntelliJ plugins, a user just needs to select a class, and tests are generated with a mouse-click.

EVOSUITE has been evaluated on millions of lines of Java code [9], both open-source code and close-source code pro-

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e	eration tool.							
	Prerequisites							
	Static or dynamic	Dynamic testing at the Java class						

Table 1: Classification of the EVOSUITE unit test gen-

Static or dynamic	Dynamic testing at the Java class level					
Software Type	Java classes					
Lifecycle phase	Unit testing for Java programs					
Environment	All Java development environments					
Knowledge required	JUnit unit testing for Java					
Experience required	Basic unit testing knowledge					
Input and Output of the tool						
Input	Bytecode of the target class and dependencies					
Output	JUnit test cases (version 3 or 4)					
Operation						
Interaction	Through the command line, and plugins for IntelliJ, Maven and Eclipse					
User guidance	manual verification of assertions for functional faults					
Source of information	http://www.evosuite.org					
Maturity	Mature research prototype, under development					
Technology behind the tool	Search-based testing / whole test suite generation					
Obtaining the tool and information						
License Lesser GPL V.3						
Cost	Open source					
Support	None					
Does there exist empirical evi	idence about					
Effectiveness and Scalability	See [6,9].					

vided by one of our industrial partners. In the first two editions of the unit testing tool competition, EVOSUITE ranked first [4,5], whereas it ranked second in the third one.

EVOSUITE uses an evolutionary approach to derive these test suites: A genetic algorithm evolves candidate individuals (chromosomes) using operators inspired by natural evolution (e.g., selection, crossover and mutation), such that iteratively better solutions with respect to the optimization target (e.g., branch coverage) are produced. For details on this test generation approach we refer to [6].

As the generated unit tests are meant for human consumption [10], EVOSUITE applies various post-processing steps to improve readability (e.g., minimising) and adds test assertions that capture the current behavior of the tested classes.

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To select the most effective assertions, EVOSUITE uses mutation analysis [11]. EVOSUITE can also be used to automatically find faults such as undeclared thrown exceptions and broken code contracts [7]. For more details on the tool and its abilities we refer to [2,3].

3. COMPETITION SETUP

EVOSUITE can be configured to target different coverage criteria. The fitness function to drive the genetic algorithm was based on a combination of several criteria [12] (e.g., line coverage, branch coverage, branch coverage by direct method invocation, weak mutation testing, output coverage, exception coverage). EVOSUITE now by default uses an archive of solutions [13], which means that throughout the search, whenever a new coverage goal is satisfied, the corresponding test is stored in the archive, and this goal is no longer targeted by the fitness function. We enabled the post-processing step of test minimization, but to reduce the time spent we included all assertions rather than filtering them with mutation analysis [11]. The use of all assertions has effects on readability and the chances of obtaining flaky tests. However, as readability is not measured by the SBST contest metric, and many of the improvements to EVOSUITE since the last competition target flaky tests, we deemed this not a problem.

In contrast to previous instances of the competition, the test generation tools this time received a time budget as input, and then had to generate tests within that time. EVOSUITE uses a combination of different timeouts for its individual phases (e.g., initialization, search, minimization, assertion generation, compilation check, removal of flaky tests), which created the challenge of distributing the overall budget onto these phases. We used a simple approach where 50% of the time was allocated to the search, whereas the other half of the time was distributed equally to the remaining phases. If any of the phases used more time than allocated, which can for example happen if test executions take long or lead to timeouts, then phases for which there is no time left are skipped. For example, if there is no time left for minimization, then the raw test suite as generated by the search is returned.

4. BENCHMARK RESULTS

The coverage results (line coverage and branch coverage) achieved by EVOSUITE on the benchmark classes are listed in Table 2. Coverage is generally in the expected range, with clear increases for higher time budgets. Coverage on several of the benchmarks from Closure is low, which matches previous findings [15].

Results in terms of mutation scores and fault detection ratio (i.e., how many of the runs had at least one failing test on the corresponding bug) are showing in Table 3. Again the Closure results are generally worse than those of other projects, both in terms of mutation score and ratio of fault detection.

On average, EVOSUITE generated 0.14 flaky tests per run, a value that is significantly lower than that of any of the competing tools. This is due to extensive efforts to isolate EVOSUITE tests from the execution environment [1]. A few of these flaky tests were introduced by recent changes to EVOSUITE following experiments on Defects4J [15]: EVO-SUITE now includes assertions on the source of exceptions, similar to commercial tools like Agitar One. Unfortunately, there were several instances in the competition where these assertions lead to flaky tests. For example, the following is an excerpt from a test for the Defects4J bug Lang-41, generated by EVOSUITE:

```
@Test(timeout = 4000)
public void test19() throws Throwable {
   Class<Double> class0 = Double.class;
   String string0 = ClassUtils.getPackageName(class0);
   try {
      ClassUtils.getClass(string0);
      fail("Expecting exception: ClassNotFoundException
               ");
   } catch(ClassNotFoundException e) {
      assertThrownBy("java.net.URLClassLoader", e);
   }
}
```

While compiling and executing this test with JUnit works without problems, the mutation analysis step of the competition used Ant to run the tests; Ant uses a complex setup of classloaders that eventually leads to the assertThrownBy in the above example to fail, as the source of the exception is a different one.

There are 15 runs in total where EVOSUITE did not produce any test suites; these are only for higher time budgets (240s, 480s). The majority of these runs are due to EVOSUITE not terminating before the hard timeout of the competition infrastructure. This may happen, for example, when test execution on the class under test takes long (e.g., timeouts), and when resetting the static state of the classes under test takes a long time. However, this number is still lower than the number of erroneous runs compared to other tools, and in the remaining 1617 runs of the competition EVOSUITE terminated in time and produced at least two tests. In 13 of these, EVOSUITE produced a test suite with a compilation error.

5. CONCLUSIONS

With an overall score of 1126.7, EVOSUITE achieved the highest score of all tools in the competition.

To learn more about EVOSUITE, visit our Web site:

http://www.evosuite.org

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Table 2: Detailed coverage 1	results of EvoSuite on	the SBST	benchmark classes.
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Defects4J ID	Class		Line Coverage						
		60s	120s	240s	480s	60s	120s	240s	480s
Chart-1	org. jfree. chart. renderer. category. AbstractCategoryItemRenderer	34.5%	36.8%	51.0%	56.2%	22.6%	26.4%	39.9%	45.7%
Chart-11	org.jfree.chart.util.ShapeUtilities	81.0%	92.1%	62.1%	79.4%	58.0%	82.8%	57.3%	76.6%
Chart-12	org.jfree.chart.plot.MultiplePiePlot	54.2%	62.0%	64.9%	51.3%	48.0%	53.4%	55.6%	45.8%
Chart-16	org.jfree.data.category.DefaultIntervalCategoryDataset	81.8%	82.7%	85.2%	88.2%	74.6%	78.1%	80.1%	85.0%
Chart-17	org.jfree.data.time.TimeSeries	69.2%	83.0%	87.3%	89.5%	58.9%	74.3%	80.9%	84.5%
Chart-2 Chart 20	org.jiree.data.generai.DatasetUtilities	0.0%	18.0%	07.2%	11.4%	0.0%	13.9%	02.2%	75.0%
Chart-20 Chart-23	org.jfree.chart.piot.valueMarker	95.5% 41.6%	95.5%	90.0% 43.1%	30.8%	30.7%	36.6%	34.8%	26.2%
Chart-25 Chart-24	org ifree chart renderer GravPaintScale	100.0%	100.0%	100.0%	100.0%	100.0%	100.0%	100.0%	100.0%
Chart-25	org ifree chart renderer category. StatisticalBarBenderer	13.1%	13.4%	18.4%	26.8%	8.3%	8.9%	10.0%	16.5%
Chart-26	org.ifree.chart.axis.Axis	54.6%	63.7%	73.5%	65.7%	37.5%	56.9%	63.2%	57.1%
Chart-3	org.jfree.data.time.TimeSeries	77.5%	78.8%	90.3%	93.2%	60.5%	64.7%	81.4%	87.5%
Chart-4	org.jfree.chart.plot.XYPlot	45.5%	48.1%	55.9%	60.8%	29.5%	32.5%	41.6%	47.6%
Chart-6	org.jfree.chart.util.ShapeList	51.0%	54.9%	55.4%	55.4%	37.5%	49.0%	49.0%	50.0%
Chart-7	org.jfree.data.time.TimePeriodValues	58.6%	95.3%	97.0%	98.1%	38.3%	89.2%	95.4%	98.5%
Chart-9	org.jfree.data.time.TimeSeries	65.8%	79.8%	83.6%	92.3%	53.3%	69.8%	75.6%	87.0%
Closure-100	com.google.javascript.jscomp.CheckGlobalThis	24.4%	42.6%	40.7%	51.1%	3.6%	21.7%	21.1%	29.2%
Closure-124	com.google.javascript.jscomp.ExploitAssigns	9.5%	12.8%	36.7%	39.9%	0.0%	2.5%	22.3%	25.7%
Closure-130	com.google.javascript.jscomp.CollapseProperties	9.1%	11.3%	13.1%	15.4%	2.6%	3.6%	5.3%	6.5%
Closure-132	com.google.javascript.jscomp.PeepnoieSubstituteAiternateSyntax	9.8%	13.4%	21.0%	14.1%	3.1%	4.8%	0.3%	8.4%
Closure-16	com google javascript.jscomp.Controir lowAnalysis	9.470 19.4%	18.3%	31.9% 94.9%	24.2%	0.0%	5.6%	20.470	20.9%
Closure-20	com google javascript iscomp PeepholeSubstituteAlternateSuntay	11.470	12.370	12 7%	17.0%	4.0%	4.7%	7.9%	10.8%
Closure-46	com google javascript rhino istyne RecordTyne	2.3%	2.3%	75.6%	91.1%	0.0%	0.0%	63.7%	75.7%
Closure-68	com.google.javascript.iscomp.parsing.JsDocInfoParser	16.6%	17.4%	22.9%	33.5%	11.3%	12.4%	16.5%	24.6%
Closure-74	com.google.javascript.jscomp.PeepholeFoldConstants	5.6%	5.8%	8.9%	10.9%	0.2%	0.2%	5.0%	6.5%
Closure-98	com.google.javascript.jscomp.ReferenceCollectingCallback	29.2%	35.3%	57.3%	63.0%	5.8%	9.0%	20.6%	26.3%
Closure-99	com.google.javascript.jscomp.CheckGlobalThis	24.4%	43.3%	42.6%	48.5%	3.2%	20.2%	19.9%	23.9%
Lang-28	org.apache.commons.lang3.text.translate.NumericEntityUnescaper	12.0%	23.3%	47.3%	75.3%	6.2%	27.1%	51.0%	79.2%
Lang-33	org.apache.commons.lang3.ClassUtils	80.7%	81.9%	84.6%	77.0%	60.9%	61.0%	66.7%	59.9%
Lang-36	org.apache.commons.lang3.math.NumberUtils	83.9%	86.2%	90.3%	92.1%	67.1%	70.8%	76.0%	80.0%
Lang-37	org.apache.commons.lang3.ArrayUtils	78.6%	83.5%	88.1%	90.1%	61.5%	66.5%	72.2%	75.1%
Lang-41	org.apache.commons.lang.ClassUtils	81.4%	84.8%	86.1%	76.2%	61.3%	65.6%	67.7%	59.2%
Lang-43	org.apache.commons.lang.text.ExtendedMessageFormat	49.6%	54.1%	53.3%	62.9%	32.0%	37.9%	41.6%	54.4%
Lang-47	org.apache.commons.lang.text.StrBuilder	75.7%	82.0%	86.2%	90.6%	63.8%	72.3%	78.7%	84.6%
Lang-50	org.apacne.commons.lang.time.rastDateFormat	80.4%	03.0% 86.407	80.1%	80.3% of 70/	75.007	43.4%	13.8%	70.1%
Lang-57	org.apache.commons.tang.LocaleUtils	70.407	00.470	04.070 96.907	00.170	64 107	10.870 GE 007	79 407	70.170
Lang-59	org apache commons lang text StrBuilder	79.470	70.0%	86.0%	89.870	68.8%	71.4%	79.1%	82.0%
Lang-60	org apache commons lang text StrBuilder	77.3%	80.9%	85.5%	88.6%	67.2%	72.9%	77.0%	81.3%
Lang-63	org apache commons lang time DurationFormatUtils	81.1%	84.8%	82.6%	92.1%	78.8%	82.6%	80.2%	90.1%
Lang-65	org.apache.commons.lang.time.DateUtils	72.5%	89.3%	94.1%	97.5%	56.6%	81.3%	87.9%	93.6%
Math-103	org.apache.commons.math.distribution.NormalDistributionImpl	85.7%	85.7%	85.7%	85.7%	77.8%	77.8%	77.8%	77.8%
Math-106	org.apache.commons.math.fraction.ProperFractionFormat	63.5%	80.2%	85.4%	79.9%	46.5%	70.2%	76.3%	70.2%
Math-18	org.apache.commons.math3.optimization.direct.CMAESOptimizer	85.7%	86.1%	82.3%	84.2%	71.0%	72.2%	66.7%	69.6%
Math-2	org.apache.commons.math3.distribution.HypergeometricDistribution	98.5%	98.2%	99.7%	99.5%	96.8%	96.2%	98.7%	99.4%
Math-20	org.apache.commons.math3.optimization.direct.CMAESOptimizer	85.4%	87.3%	86.7%	87.3%	67.7%	71.8%	71.9%	72.9%
Math-21	org.apache.commons.math3.linear.RectangularCholeskyDecomposition	58.5%	69.7%	73.3%	91.5%	58.3%	70.2%	73.2%	91.1%
Math-39	org.apache.commons.math.ode.nonstiff.EmbeddedRungeKuttaIntegrator	74.4%	67.1%	89.5%	90.9%	39.2%	36.1%	69.1%	71.0%
Math-44	org.apache.commons.math.ode.AbstractIntegrator	63.1%	62.6%	68.2%	73.3%	49.6%	50.7%	58.3%	64.5%
Math-52	org.apache.commons.math.geometry.euclidean.threed.Rotation	72.9%	81.3%	97.0%	98.7%	45.5%	59.0%	88.7%	91.5%
Math-56	org.apache.commons.math.util.MultidimensionalCounter	97.9%	99.3%	99.8%	99.8%	94.3%	95.8%	98.4%	98.4%
Math-64	org.apache.commons.math.optimization.general.LevenbergMarquardtOptimizer	35.8%	40.9%	66.5%	76.9%	22.7%	27.6%	55.0%	64.8%
Math-67	org.apacne.commons.math.optimization.MultiStartUnivariateRealOptimizer	87.0%	88.1%	80.9%	90.0%	81.9%	90.7%	50.8%	97.1%
Math 88	org apache commons math optimization linear Simpley Tableau	SO.076 S 407	00.970 84.0%	04.2%	06.1%	29.0%	39.470 74.9%	99.270 86.0%	05.870
Math-00	org apache commons math fraction Fraction	96.5%	07.9%	94.370	90.170	0/1 1%	95.9%	97.4%	90.170
Math-91 Math-93	org apache commons math util MathIltils	81.6%	84.1%	86.0%	87.5%	78.9%	80.4%	83.5%	85.9%
Time-10	org.ioda.time.base.BaseSingleFieldPeriod	91.4%	91.4%	88.6%	91.2%	87.3%	90.6%	86.6%	92.4%
Time-11	org.joda.time.tz.ZoneInfoCompiler	30.7%	32.5%	46.3%	34.0%	25.9%	27.9%	40.2%	30.7%
Time-13	org.joda.time.format.PeriodFormatterBuilder	49.8%	52.7%	65.5%	72.7%	29.0%	31.0%	44.6%	54.5%
Time-20	org.joda.time.format.DateTimeFormatterBuilder	35.3%	36.5%	68.5%	77.0%	25.3%	25.5%	52.6%	60.9%
Time-23	org.joda.time.DateTimeZone	83.6%	83.9%	84.0%	85.4%	71.5%	72.7%	73.9%	77.5%
Time-3	org.joda.time.MutableDateTime	87.4%	91.0%	93.6%	95.4%	79.6%	84.4%	90.5%	96.8%
Time-4	org.joda.time.Partial	72.5%	77.0%	93.5%	97.8%	55.5%	61.0%	87.1%	93.6%
Time-5	org.joda.time.Period	96.7%	97.3%	98.4%	98.9%	81.0%	84.4%	90.6%	93.2%
Time-7	org.joda.time.format.DateTimeFormatter	87.3%	86.8%	90.1%	91.7%	84.1%	80.3%	84.1%	87.5%
Time-8	org.joda.time.DateTimeZone	81.3%	81.4%	82.8%	84.8%	68.3%	70.4%	72.9%	77.7%
Average		58.3%	63.7%	71.1%	73.9%	46.6%	53.5%	61.7%	65.6%

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Table 3: Detailed fault detection results of EVOSUITE on the SBST benchmark classes.

Defects4J ID	Class	Mutation Score Fault Detec		etection					
		60s	120s	240s	480s	60s	120s	240s	480s
		11.004	10.00		a.a. 104	a a64	0.004		0.00
Chart-1	org.jfree.chart.renderer.category.AbstractCategoryItemRenderer	11.3%	12.2%	17.4%	22.4%	0.0%	0.0%	16.7%	0.0%
Chart-11	org.jfree.chart.util.ShapeUtilities	13.4%	28.0%	18.9%	25.4%	50.0%	100.0%	66.7%	83.3%
Chart-12	org.jfree.chart.plot.MultiplePiePlot	10.8%	19.6%	20.3%	17.7%	16.7%	16.7%	0.0%	0.0%
Chart-16	org ifree data category DefaultIntervalCategoryDataset	47.3%	48.6%	52.7%	55.4%	100.0%	100.0%	100.0%	100.0%
Chart 17		27.0%	20.70/	20.0%	49.407	100.0%	100.0%	100.0%	100.0%
Chart-17	org. free. data. time. Timeseries	21.970	32.170	39.0%	42.470	100.0%	100.0%	100.0%	100.0%
Chart-2	org.jfree.data.general.DatasetUtilities	0.0%	7.3%	43.4%	49.9%	0.0%	0.0%	0.0%	0.0%
Chart-20	org.jfree.chart.plot.ValueMarker	81.8%	81.8%	81.8%	69.7%	0.0%	0.0%	0.0%	0.0%
Chart-23	org.ifree.chart.renderer.category.MinMaxCategoryRenderer	3.4%	8.4%	6.9%	6.3%	0.0%	83.3%	50.0%	66.7%
Chart-24	org ifree chart renderer GravPaintScale	91.9%	65.3%	73.4%	82.9%	100.0%	33 3%	50.0%	66.7%
Clark 05	orgine characterider of any antibodie in De Dechara	0.207	1.007	0.407	02.070	0.0%	0.007	50.0%	50.007
Chart-25	org. Inte. chart. renderer. category. Statistical barkenderer	0.3%	1.970	2.470	2.370	0.0%	0.0%	30.0%	30.0%
Chart-26	org.jfree.chart.axis.Axis	5.7%	17.1%	19.8%	17.5%	0.0%	0.0%	16.7%	0.0%
Chart-3	org.jfree.data.time.TimeSeries	28.0%	33.8%	44.9%	47.8%	0.0%	16.7%	33.3%	50.0%
Chart-4	org.ifree.chart.plot.XYPlot	17.3%	17.2%	19.8%	36.6%	16.7%	0.0%	100.0%	100.0%
Chart-6	org ifree chart util ShapeList	15.5%	23.8%	19.0%	20.2%	83 3%	100.0%	83 3%	100.0%
Chart 7	and if the data time. Time Darie dValues	01.70/	62.007	67.907	61.907	0.0%	16 707	E0.0%	CC 70/
Chart-7	org. inter-cata. time. I mer eriod values	21.170	03.970	07.370	01.270	0.0%	10.770	30.0%	00.770
Chart-9	org.jhree.data.time.TimeSeries	23.0%	31.5%	36.2%	44.1%	0.0%	16.7%	0.0%	0.0%
Closure-100	com.google.javascript.jscomp.CheckGlobalThis	8.2%	18.2%	20.6%	23.6%	0.0%	0.0%	0.0%	16.7%
Closure-124	com.google.javascript.jscomp.ExploitAssigns	1.4%	2.5%	15.1%	16.9%	0.0%	0.0%	0.0%	0.0%
Closure-130	com google isvescript iscomp CollepseProperties	0.0%	0.4%	0.7%	1.3%	0.0%	0.0%	0.0%	0.0%
Closure-100	configuration of the second seco	0.070	0.470	7 407	12.70	0.070	0.070	0.070	0.0%
Closure-152	com.google.javascript.jscomp.reepinoieSubstituteAnternateSyntax	3.370	0.070	1.470	13.770	0.0%	0.0%	0.0%	0.0%
Closure-14	com.google.javascript.jscomp.ControlFlowAnalysis	2.7%	3.6%	20.0%	26.3%	0.0%	0.0%	0.0%	0.0%
Closure-16	com.google.javascript.jscomp.ScopedAliases	0.0%	4.0%	5.6%	4.4%	0.0%	0.0%	0.0%	0.0%
Closure-20	com.google.javascript.jscomp.PeepholeSubstituteAlternateSyntax	3.9%	8.5%	10.0%	13.4%	0.0%	0.0%	0.0%	0.0%
Closure-46	com google javascrint rhino istyne BecordTyne	0.0%	0.0%	37 4%	43.1%	0.0%	0.0%	16.7%	0.0%
Closure-40	configuration of the second state of the secon	0.070	4 407	0.207	10.170	0.070	0.070	0.007	0.070
Closure-68	com.googie.javascript.jscomp.parsing.jsDocinioParser	3.1%	4.4%	9.3%	13.5%	0.0%	0.0%	0.0%	0.0%
Closure-74	com.google.javascript.jscomp.PeepholeFoldConstants	1.9%	2.0%	2.5%	3.2%	0.0%	0.0%	0.0%	0.0%
Closure-98	com.google.javascript.jscomp.ReferenceCollectingCallback	8.3%	9.1%	22.4%	27.6%	0.0%	0.0%	0.0%	0.0%
Closure-99	com.google.javascript.jscomp.CheckGlobalThis	7.0%	17.0%	18.2%	21.1%	0.0%	0.0%	0.0%	0.0%
Lang-28	org apache commons lang3 text translate NumericEntityUnescaper	8 3%	17.7%	24.2%	40.8%	0.0%	0.0%	0.0%	0.0%
Lang-20	organization of the second s	26.007	20.407	49.70	40.070	100.007	100.007	100.007	0.070
Lang-33	org.apacne.commons.lang3.ClassUtils	30.2%	39.4%	43.170	38.1%	100.0%	100.0%	100.0%	83.370
Lang-36	org.apache.commons.lang3.math.NumberUtils	27.1%	36.2%	38.7%	41.8%	16.7%	0.0%	83.3%	83.3%
Lang-37	org.apache.commons.lang3.ArrayUtils	24.4%	35.0%	56.6%	58.2%	16.7%	0.0%	0.0%	33.3%
Lang-41	org.apache.commons.lang.ClassUtils	39.1%	44.2%	46.5%	36.8%	100.0%	100.0%	100.0%	100.0%
Long 42	org angle common long tort Extended Message Format	19.4%	16.6%	10.4%	25.8%	100.0%	100.0%	100.0%	100.0%
Lang-40	org.apache.commons.tang.text.Extendedivessagerormat	12.4/0	10.076	13.470	20.070	100.076	100.076	100.076	100.070
Lang-47	org.apache.commons.lang.text.StrBuilder	23.2%	32.2%	44.0%	54.6%	66.7%	66.7%	66.7%	83.3%
Lang-50	org.apache.commons.lang.time.FastDateFormat	19.2%	22.9%	50.8%	56.5%	0.0%	0.0%	0.0%	16.7%
Lang-57	org.apache.commons.lang.LocaleUtils	40.5%	39.2%	37.7%	38.4%	83.3%	100.0%	100.0%	100.0%
Lang=58	org anache commons lang math NumberUtils	21.1%	33.3%	37.6%	42.5%	0.0%	0.0%	0.0%	0.0%
Long 50	org apache commons lang tart StrBuilder	25 20%	21.1%	11 80%	40.0%	16.7%	22 20%	82.2%	66 7%
Lang=0.9	org.apache.commons.lang.text.Str Duride	20.070	01.170	44.070	43.370	10.770	10.5%	10.570	00.170
Lang-60	org.apache.commons.lang.text.StrBuilder	23.8%	34.0%	44.9%	48.6%	16.7%	16.7%	16.7%	33.3%
Lang-63	org.apache.commons.lang.time.DurationFormatUtils	13.6%	20.4%	20.8%	25.5%	0.0%	0.0%	0.0%	16.7%
Lang-65	org.apache.commons.lang.time.DateUtils	28.1%	35.7%	51.9%	55.3%	33.3%	16.7%	83.3%	100.0%
Math-103	org anache commons math distribution NormalDistributionImpl	63.0%	62.8%	63.7%	63.0%	0.0%	0.0%	0.0%	0.0%
Math 106	one and the commons moth fraction Date of the stick Format	05.0%	22.070	40.0%	25 107	0.0%	0.0%	16 707	0.0%
Math=100	org.apache.commons.math.math.interpretation.com/orproduct	20.070	02.270	40.570	33.170	0.076	0.070	10.170	0.070
Math-18	org.apache.commons.math3.optimization.direct.CMAESOptimizer	29.9%	29.0%	26.5%	26.5%	0.0%	0.0%	0.0%	0.0%
Math-2	org.apache.commons.math3.distribution.HypergeometricDistribution	39.3%	84.8%	89.5%	90.8%	0.0%	16.7%	0.0%	0.0%
Math-20	org.apache.commons.math3.optimization.direct.CMAESOptimizer	33.1%	35.9%	28.7%	27.3%	0.0%	0.0%	0.0%	0.0%
Math-21	org.apache.commons.math3.linear.BectangularCholeskyDecomposition	23.7%	32.5%	28.1%	44.0%	0.0%	0.0%	0.0%	0.0%
Math-39	org apache commons math ode nonstiff EmbeddedBungeKutteIntegrator	20.7%	17.9%	30.3%	35.6%	0.0%	0.0%	0.0%	0.0%
Math-05	organization of the second s	25.170	20.007	25.007	41 407	0.070	0.070	0.070	0.070
watn-44	org.apacne.commons.matn.ode.Abstractintegrator	35.0%	32.0%	35.0%	41.4%	0.0%	0.0%	0.0%	0.0%
Math-52	org.apache.commons.math.geometry.euclidean.threed.Rotation	52.8%	58.8%	79.5%	82.8%	0.0%	0.0%	0.0%	0.0%
Math-56	org.apache.commons.math.util.MultidimensionalCounter	48.8%	56.2%	74.3%	74.4%	16.7%	16.7%	83.3%	100.0%
Math-64	org.apache.commons.math.optimization.general.LevenbergMarquardtOptimizer	5.7%	5.9%	16.4%	24.0%	0.0%	66.7%	66.7%	83.3%
Moth 67	org angle commons math optimization MultiStart University Pool Optimizar	26.9%	52.2%	47 4%	50.4%	100.0%	100.0%	100.0%	100.0%
Math-07	org.apache.commons.math.optimization.wrutistartoinwariatertearoptimizer	00.270 00.007	20.507	41.470	03.470	0.0%	0.0%	0.0%	100.070
Math-7	org.apache.commons.math3.ode.AbstractIntegrator	26.2%	32.5%	33.1%	31.5%	0.0%	0.0%	0.0%	0.0%
Math-88	org.apache.commons.math.optimization.linear.SimplexTableau	0.5%	41.6%	52.8%	57.1%	0.0%	0.0%	16.7%	0.0%
Math-91	org.apache.commons.math.fraction.Fraction	58.7%	69.1%	70.8%	70.0%	0.0%	0.0%	0.0%	0.0%
Math-93	org.apache.commons.math.util.MathUtils	24.6%	46.0%	65.7%	68.8%	0.0%	16.7%	0.0%	16.7%
Time 10	org inde time have BayeSingleFieldPoried	46.2%	62 2%	50.6%	62 4%	0.0%	0.0%	16 7%	16 7%
Time=10	org.joda.emic.base.base.base.bilgier ieldi eriod	11 407	15 707	10.40%	14.007	100.0%	100.0%	100.007	20.170
11me-11	org.joua.time.tz.ZoneinioCompiler	11.4%	10.7%	19.4%	14.0%	100.0%	100.0%	100.0%	00.7%
Time-13	org.joda.time.tormat.PeriodFormatterBuilder	11.9%	16.8%	25.1%	30.9%	0.0%	0.0%	0.0%	16.7%
Time-20	org.joda.time.format.DateTimeFormatterBuilder	9.3%	11.2%	27.7%	39.4%	0.0%	0.0%	0.0%	0.0%
Time-23	org.joda.time.DateTimeZone	40.4%	49.3%	50.2%	51.7%	0.0%	16.7%	0.0%	0.0%
Time-3	org ioda time MutableDateTime	35 40%	47.6%	55.9%	57.6%	100.0%	50.0%	22.20%	16.7%
Time 4	org.joua.enne.wiueabiebatei inte	00.470	30 70/	67 007	79.007	16 70/	0.0%	50.0%	20.170
1 ime-4	org.joda.time.Partial	29.6%	38.7%	07.9%	13.9%	16.7%	0.0%	50.0%	33.3%
Time-5	org.joda.time.Period	25.1%	45.1%	73.4%	82.2%	33.3%	50.0%	83.3%	100.0%
Time-7	org.joda.time.format.DateTimeFormatter	39.6%	46.1%	54.6%	59.3%	33.3%	16.7%	16.7%	0.0%
Time-8	org.joda.time.DateTimeZone	33.8%	47.6%	51.6%	53.8%	66.7%	83.3%	66.7%	100.0%
	0.9								
Average		23.4%	30.6%	37.7%	41.0%	21.8%	24.3%	31.1%	33.3%

Testing (SBST), 2016.

[15] S. Shamshiri, R. Just, J. M. Rojas, G. Fraser, P. McMinn, and A. Arcuri, "Do automatically generated unit tests find real faults? an empirical study of effectiveness and challenges," in *IEEE/ACM Int. Conference on Automated Software Engineering* (ASE). IEEE, 2015, pp. 201–211.