

# EvoSUITE at the SBST 2019 Tool Competition

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## ABSTRACT

EvoSUITE is a search-based tool that automatically generates executable unit tests for Java code (JUnit tests). This paper summarises the results and experiences of EvoSUITE’s participation at the seventh unit testing competition at SBST 2019, where EvoSUITE achieved the highest overall score (255.43 points) for the sixth time in seven editions of the competition.

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## 1 INTRODUCTION

The annual unit test generation competition aims to drive and evaluate progress on unit test generation tools. In the 7th instance of the competition at the International Workshop on Search-Based Software Testing (SBST) 2019, several JUnit test generation tools competed on a set of 38 open-source Java classes. This paper describes the results obtained by the EvoSUITE test generation tool [8] in this competition. Details about the procedure of the competition, the technical framework, and the benchmark classes can be found in [21]. In this competition, EvoSUITE achieved an overall score of 255.43, which was the highest among the competing and baseline tools.

## 2 ABOUT EVOSUITE

EvoSUITE [8] is a search-based tool [12] that uses a genetic algorithm to automatically generate test suites for Java classes. Given the name of a target class and the full Java classpath (i.e., where to find the compiled bytecode of the class under test and all its dependencies), EvoSUITE automatically produces a set of JUnit test cases aimed at maximising code coverage. EvoSUITE can be used on the command line, or through plugins for popular development tools such as IntelliJ, Eclipse, or Maven [2].

**[TODO: introduce DynaMOSA]** The underlying genetic algorithm uses test suites as representation (chromosomes). Each test suite consists of a variable number of test cases, each of which is represented as a variable-length sequence of Java statements

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**Table 1: Classification of the EvoSUITE unit test generation tool**

Prerequisites	
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 4 test cases
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	<a href="http://www.evospace.org">http://www.evospace.org</a>
Maturity	Mature research prototype, under development
Technology behind the tool	Search-based testing / whole test suite generation <b>[TODO: not whole test suite anymore]</b>
Obtaining the tool and information	
License	Lesser GPL V.3
Cost	Open source
Support	None
Does there exist empirical evidence about	
Effectiveness and Scalability	See [12, 13]

(e.g., calls on the class under test). A population of randomly generated individuals is evolved using suitable search operators (e.g., selection, crossover and mutation), such that iteratively better solutions with respect to the optimisation target are produced. The optimisation target is to maximise code coverage. To achieve this, the fitness function uses standard heuristics such as the branch distance; see [12] for more details. EvoSUITE can be configured to optimise for multiple coverage criteria at the same time, and the default configuration combines branch coverage with mutation testing [14] and other basic criteria [22]. Once the search is completed, EvoSUITE applies various optimisations to improve the readability of the generated tests. For example, tests are minimised,

and a minimised set of effective test assertions is selected using mutation analysis [18]. For more details on the tool and its abilities we refer to [8, 9].

The effectiveness of EvoSUITE has been evaluated on open source as well as industrial software in terms of code coverage [7, 13, 24], fault finding effectiveness [1, 26], and effects on developer productivity [17, 23] and software maintenance [27].

EvoSUITE has a longstanding record of success at the unit testing tool competition, having ranked second in the third edition of the competition [15] and first in all the other editions [10, 11, 16, 19, 20].

### 3 COMPETITION SETUP

The configuration of EvoSUITE for the 2019 competition is largely based on its default values, since these have been tuned extensively [4]. We used the default set of coverage criteria [22] (e.g., line coverage, branch coverage, branch coverage by direct method invocation, weak mutation testing, output coverage, exception coverage). EvoSUITE uses an archive of solutions [24] to keep the search focused on uncovered goals, iteratively discarding covered goals and storing the tests that covered them. After a certain percentage of the search budget has passed and with a certain probability, EvoSUITE starts using mock objects (relying on Mockito) instead of actual class instances [3] for dependencies; only branches that cannot be covered without mocks lead to tests with mock objects. As in the previous instance of the competition, we continue to use frequency-based weighted constants for seeding [25] and support Java Enterprise Edition features [5]. EvoSUITE is actively maintained, therefore several bug fixes and minor improvements have been applied since the last instance of the competition, in particular in relation to non-determinism and flaky tests [6], as well as new types of test assertions (e.g., related to arrays and standard container classes).

Like in previous instances of the competition, we enabled the post-processing step of test minimisation—not for efficiency reasons, but because minimised tests are less likely to break. To reduce the overall time of test generation we included all assertions rather than filtering them with mutation analysis [18], which is computationally expensive. The use of all assertions has a negative impact on readability, but this is not evaluated as part of the SBST contest.

Four time budgets were used to call each tool: 10, 60, 120 and 240 seconds. We used the same strategy used in previous competition (e.g., [16]) to distribute the overall time budget onto the different phases of EvoSUITE (e.g., initialisation, search, minimisation, assertion generation, compilation check, removal of flaky tests). That is, 50% of the time was allocated to the search, and the rest was distributed equally to the remaining phases.

### 4 BENCHMARK RESULTS

*Overall Results.* [TODO: ]

*Flaky Tests.* [TODO: ]

*Problematic benchmarks.* [TODO: why did we get 0% coverage on most AUTHZFORCE benchmarks? and SPOON-169, SPOON-25?] [18]

[TODO: why did we get 0% coverage on DUBBO-2 and WEBMAGIC-4? classpath was incorrect.]

[TODO: what did we get 0% coverage on SPOON-253, SPOON-20, SPOON-16, SPOON-155 for larger search budgets?]

[TODO: why did not we get data for FASTJSON-1, FASTJSON-3, FESCAR-41, FESCAR-42, SPOON-105 for a search budget of 10 seconds?]

### 5 CONCLUSIONS

This paper reports on the participation of the EvoSUITE test generation tool in the 7th SBST Java Unit Testing Tool Contest. With an overall score of 255.43 points, EvoSUITE achieved the highest score of all tools in the competition.

To learn more about EvoSUITE, visit our Web site:

<http://www.ev-suite.org>

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### REFERENCES

- [1] M. M. Almasi, H. Hemmati, G. Fraser, A. Arcuri, and J. Benefelds, “An industrial evaluation of unit test generation: Finding real faults in a financial application,” in *ACM/IEEE Int. Conference on Software Engineering (ICSE)*. IEEE, 2017, pp. 263–272.
- [2] A. Arcuri, J. Campos, and G. Fraser, “Unit Test Generation During Software Development: EvoSuite Plugins for Maven, IntelliJ and Jenkins,” in *IEEE International Conference on Software Testing, Verification and Validation (ICST)*. IEEE Computer Society, 2016, pp. 401–408.
- [3] A. Arcuri, G. Fraser, and R. Just, “Private api access and functional mocking in automated unit test generation,” in *IEEE Int. Conference on Software Testing, Verification and Validation (ICST)*, 2017, pp. 126–137.
- [4] A. Arcuri and G. Fraser, “Parameter tuning or default values? An empirical investigation in search-based software engineering,” *Empirical Software Engineering (EMSE)*, pp. 1–30, 2013.
- [5] —, “Java enterprise edition support in search-based junit test generation,” in *Int. Symposium on Search Based Software Engineering*. Springer, 2016, pp. 3–17.
- [6] A. Arcuri, G. Fraser, and J. P. Galeotti, “Automated unit test generation for classes with environment dependencies,” in *IEEE/ACM Int. Conference on Automated Software Engineering (ASE)*. ACM, 2014, pp. 79–90.
- [7] J. Campos, Y. Ge, N. Albunian, G. Fraser, M. Eler, and A. Arcuri, “An empirical evaluation of evolutionary algorithms for unit test suite generation,” *Information and Software Technology*, vol. 104, pp. 207–235, 2018.
- [8] G. Fraser and A. Arcuri, “EvoSuite: Automatic test suite generation for object-oriented software,” in *ACM Symposium on the Foundations of Software Engineering (FSE)*, 2011, pp. 416–419.
- [9] —, “EvoSuite: On the challenges of test case generation in the real world (tool paper),” in *IEEE Int. Conference on Software Testing, Verification and Validation (ICST)*, 2013, pp. 362–369.
- [10] —, “EvoSuite at the SBST 2013 tool competition,” in *Int. Workshop on Search-Based Software Testing (SBST)*, 2013, pp. 406–409.
- [11] —, “EvoSuite at the second unit testing tool competition,” in *Fittest Workshop*. Springer, 2013, pp. 95–100.
- [12] —, “Whole test suite generation,” *IEEE Transactions on Software Engineering*, vol. 39, no. 2, pp. 276–291, 2013.
- [13] —, “A large-scale evaluation of automated unit test generation using EvoSuite,” *ACM Transactions on Software Engineering and Methodology (TOSEM)*, vol. 24, no. 2, pp. 8:1–8:42, 2014.
- [14] —, “Achieving scalable mutation-based generation of whole test suites,” *Empirical Software Engineering (EMSE)*, vol. 20, no. 3, pp. 783–812, 2015.
- [15] —, “EvoSuite at the SBST 2015 tool competition,” in *Int. Workshop on Search-Based Software Testing (SBST)*. IEEE Press, 2015, pp. 25–27.
- [16] —, “EvoSuite at the SBST 2016 tool competition,” in *Int. Workshop on Search-Based Software Testing (SBST)*. ACM, 2016, pp. 33–36.
- [17] G. Fraser, M. Staats, P. McMinn, A. Arcuri, and F. Padberg, “Does automated unit test generation really help software testers? a controlled empirical study,” *ACM Transactions on Software Engineering and Methodology (TOSEM)*, vol. 24, no. 4, pp. 23:1–23:49, 2015.
- [18] G. Fraser and A. Zeller, “Mutation-driven generation of unit tests and oracles,” *IEEE Transactions on Software Engineering (TSE)*, vol. 28, no. 2, pp. 278–292, 2012.

**Table 2: Detailed results of EvoSUITE on the SBST benchmark classes.**

Benchmark	Java Class	Branch Coverage				Mutation Score			
		10s	60s	120s	240s	10s	60s	120s	240s
AUTHZFORCE-1	org.ow2.authzforce.core.pdp.impl.PdpBean	0.0%	16.7%	0.0%	0.0%	0.0%	14.4%	0.0%	0.0%
AUTHZFORCE-11	org.ow2.authzforce.core.pdp.impl.func.LogicalNoFunction	0.0%	0.0%	0.0%	0.0%	4.8%	4.8%	0.0%	0.0%
AUTHZFORCE-27	org.ow2.authzforce.core.pdp.impl.func.MapFunctionFactory	75.0%	0.0%	0.0%	0.0%	75.0%	0.0%	0.0%	0.0%
AUTHZFORCE-32	org.ow2.authzforce.core.pdp.impl.func.SubstringFunction	0.0%	0.0%	0.0%	0.0%	7.7%	0.0%	0.0%	0.0%
AUTHZFORCE-33	org.ow2.authzforce.core.pdp.impl.SchemaHandler	55.6%	0.0%	0.0%	0.0%	57.6%	0.0%	0.0%	0.0%
AUTHZFORCE-48	org.ow2.authzforce.core.pdp.impl.policy.PolicyVersions	0.0%	12.1%	34.8%	60.6%	0.0%	16.7%	37.3%	69.6%
AUTHZFORCE-5	org.ow2.authzforce.core.pdp.impl.CloseableAttributeProvider	20.5%	0.0%	0.0%	0.0%	27.4%	0.0%	0.0%	0.0%
AUTHZFORCE-52	org.ow2.authzforce.core.pdp.impl.PepActionExpression	91.7%	0.0%	0.0%	0.0%	100.0%	0.0%	0.0%	0.0%
AUTHZFORCE-63	org.ow2.authzforce.core.pdp.impl.combining.DPUnlessPDCombiningAlg	10.0%	3.3%	0.0%	0.0%	5.6%	1.9%	0.0%	0.0%
AUTHZFORCE-65	org.ow2.authzforce.core.pdp.impl.combining.FirstApplicableCombiningAlg	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
DUBBO-2	com.alibaba.dubbo.common.utils.ReflectUtils	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
FASTJSON-1	com.alibaba.fastjson.parser.JSONLexerBase	-	23.0%	24.9%	31.6%	-	7.7%	8.8%	6.5%
FASTJSON-3	com.alibaba.fastjson.parser.DefaultJSONParser	-	13.4%	20.5%	25.7%	-	9.4%	14.1%	19.1%
FESCAR-1	com.alibaba.fescar.core.protocol.transaction.BranchReportRequest	97.9%	95.8%	96.9%	99.0%	97.7%	97.7%	98.5%	100.0%
FESCAR-12	com.alibaba.fescar.core.rpc.netty.RpcServerHandler	12.5%	87.5%	87.5%	87.5%	50.0%	100.0%	100.0%	100.0%
FESCAR-18	com.alibaba.fescar.core.protocol.MergedWarpMessage	61.9%	48.8%	56.0%	42.9%	81.2%	60.9%	68.1%	51.4%
FESCAR-23	com.alibaba.fescar.core.protocol.MergeResultMessage	63.1%	78.6%	64.3%	83.3%	78.6%	97.0%	79.2%	98.8%
FESCAR-25	com.alibaba.fescar.core.rpc.netty.RmMessageListener	37.5%	62.5%	62.5%	50.0%	22.2%	22.2%	22.2%	18.5%
FESCAR-36	com.alibaba.fescar.core.protocol.RegisterRMRequest	100.0%	92.9%	80.1%	83.3%	100.0%	93.8%	81.2%	83.0%
FESCAR-37	com.alibaba.fescar.core.rpc.RpcContext	52.9%	89.2%	91.2%	91.2%	71.9%	96.4%	94.8%	95.8%
FESCAR-41	com.alibaba.fescar.core.rpc.netty.RmRpcClient	-	2.0%	2.0%	2.0%	-	2.4%	2.4%	2.4%
FESCAR-42	com.alibaba.fescar.core.rpc.DefaultServerMessageListenerImpl	-	16.1%	27.4%	27.9%	-	16.7%	27.8%	29.2%
FESCAR-7	com.alibaba.fescar.core.rpc.netty.MessageCodecHandler	37.8%	78.4%	82.8%	87.1%	37.9%	92.8%	95.2%	98.3%
OKIO-1	okio.Buffer	27.4%	57.0%	62.3%	67.9%	13.2%	20.2%	16.3%	18.5%
OKIO-4	okio.RealBufferedSource	19.2%	37.6%	51.3%	76.2%	13.1%	28.8%	40.1%	45.3%
SPOON-105	spoon.support.compiler.jdt.PositionBuilder	-	22.1%	16.4%	3.8%	-	27.7%	23.1%	5.4%
SPOON-155	spoon.reflect.visitor.filter.AllMethodsSameSignatureFunction	12.0%	4.2%	2.5%	0.0%	19.0%	10.7%	4.3%	5.0%
SPOON-16	spoon.reflect.path.CtElementPathBuilder	34.0%	34.3%	0.0%	2.0%	45.5%	43.9%	0.0%	4.0%
SPOON-169	spoon.reflect.visitor.ImportScannerImpl	-	0.0%	0.0%	0.0%	-	0.0%	0.0%	0.0%
SPOON-20	spoon.support.reflect.reference.CtLocalVariableReferenceImpl	33.3%	20.0%	0.0%	0.0%	40.3%	33.3%	0.0%	0.0%
SPOON-211	spoon.reflect.path.impl.CtRolePathElement	8.1%	19.2%	0.0%	5.1%	27.8%	36.8%	0.0%	17.7%
SPOON-25	spoon.pattern.internal.ValueConvertorImpl	-	-	0.0%	0.0%	-	-	0.0%	0.0%
SPOON-253	spoon.pattern.internal.parameter.MapParameterInfo	29.7%	50.0%	0.0%	0.0%	34.8%	50.0%	0.0%	0.0%
SPOON-32	spoon.MavenLauncher	12.5%	10.4%	10.4%	12.5%	6.7%	5.6%	5.6%	6.7%
SPOON-65	spoon.support.DefaultCoreFactory	9.0%	49.8%	20.0%	5.7%	1.2%	2.3%	1.2%	2.0%
WEBMAGIC-1	us.codecraft.webmagic.model.PageModelExtractor	28.0%	27.6%	28.8%	31.3%	50.4%	50.0%	51.3%	54.3%
WEBMAGIC-4	us.codecraft.webmagic.Page	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
ZXING-10	com.google.zxing.qrcode.encoder.Encoder	70.1%	82.3%	82.9%	84.0%	63.0%	76.1%	72.6%	72.8%
Average		17.3%	17.2%	14.8%	15.5%	19.9%	16.9%	13.9%	14.6%

- [19] J. C. Gordon Fraser, José Miguel Rojas and A. Arcuri, "EvoSuite at the SBST 2017 tool competition," in *Int. Workshop on Search-Based Software Testing (SBST)*. IEEE Press, 2017, pp. 39–41.
- [20] J. M. R. Gordon Fraser and A. Arcuri, "EvoSuite at the SBST 2018 tool competition," in *Int. Workshop on Search-Based Software Testing (SBST)*. IEEE Press, 2018, pp. 34–37.
- [21] F. Kifetew, X. Devroey, and U. Rueda, "Java Unit Testing Tool Competition - Seventh Round," in *Int. Workshop on Search-Based Software Testing (SBST)*, 2019.
- [22] J. M. Rojas, J. Campos, M. Vivanti, G. Fraser, and A. Arcuri, "Combining multiple coverage criteria in search-based unit test generation," in *Search-Based Software Engineering*. Springer, 2015, pp. 93–108.
- [23] J. M. Rojas, G. Fraser, and A. Arcuri, "Automated unit test generation during software development: A controlled experiment and think-aloud observations," in *ACM Int. Symposium on Software Testing and Analysis (ISSTA)*. ACM, 2015, pp. 338–349.
- [24] J. M. Rojas, M. Vivanti, A. Arcuri, and G. Fraser, "A Detailed Investigation of the Effectiveness of Whole Test Suite Generation," *Empirical Software Engineering (EMSE)*, vol. 22, no. 2, pp. 852–893, 2016.
- [25] A. Sakti, G. Pesant, and Y.-G. Guéhenec, "Instance generator and problem representation to improve object oriented code coverage," *IEEE Transactions on Software Engineering*, vol. 41, no. 3, pp. 294–313, 2015.
- [26] 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.
- [27] S. Shamshiri, J. M. Rojas, J. P. Galeotti, N. Walkinshaw, and G. Fraser, "How do automatically generated unit tests influence software maintenance?" in *IEEE Int.*

*Conference on Software Testing, Verification and Validation (ICST)*, 2018, to appear.