



Software Engineering

Landscape depending parameter tuning for search-based software testing

Overview

- 1. Introduction
- 2. Fundamentals
- 3. Experimental
- 4. Adaptive parameter control
- 5. Evaluation
- 6. Conclusion
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Introduction

- Unit tests
- maximize coverage (line, branch, exception)
- lack of sufficient tests
- costly and time-consuming
- => use search-based software testing

Motivation

- Tools... => EvoSuite state-of-the-art
- may not terminate => search budget
- optimal only with optimal configuration
- No Free Lunch theorem
 - impossible to find optimal configuration for all problems
- EvoSuite's default configuration is fairly good, but not perfect

Research goal

- wide variety of problem-cases
- concept landscape depending
- adaptive parameter control
- increase coverage of EvoSuite

State-of-the-art



Challenges



Delimitation

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Search-based software testing

- tests for object oriented languages
- sequence of method calls

Fitness function

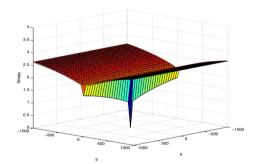
- function for e.g. coverage
- guidance for search algorithms

Fitness landscape



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Genetic algorithm

Heading

- start with random population
- iterate till termination condition
- return last generation

Heading

DynaMOSA

Heading

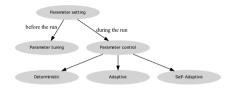
- start with random population
- multiple target
- keep track of target covering individuals

Heading

- iterate till termination condition
- update targets
- update archive
- select by rank
- return archive as last generation

Parameter tuning and control





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Corpus

SF110

- 110 open-source Java projects
- 23,894 Java Classes

Panichella et al.

- 117 open-source Java projects
- 346 Java Classes
- non-trivial and complex
- often used

Prediction sample

- S₁
- 709 Java Classes
- randomly selected
- SF110

- 9.8 days on three machines

Evaluation sample

- S_2
- 346 Java Classes
- the whole Panichella corpus

- 4.8 days on tree machines

Sensitive sample

- $\blacksquare S_3$
- 20 Java Classes
- extracted from S₁
- high Standard Deviation

- 6.6 hours on tree machines

Comparisons

- 30 repeats for every Java Class
- Mann-Whitney U-test
- Vargha-Delaney effect size Â₁₂

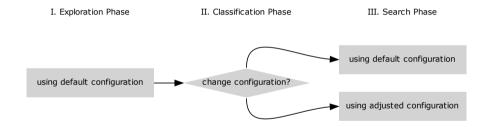
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Concept

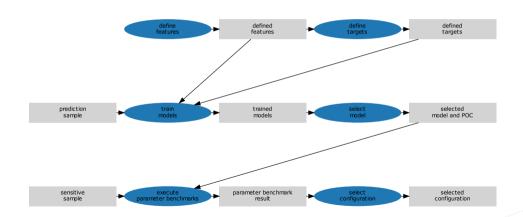
- use the beginning of the search to explore the search space
- classify the current search ("adjust" or "default")
- use configuration depending on class



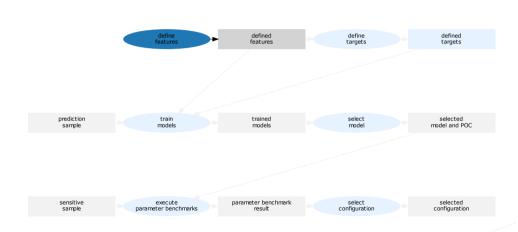
Configuration selection



Components



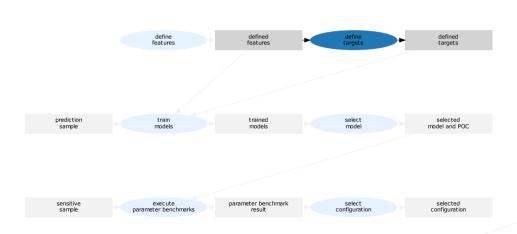
Landscape features



Landscape features

- Fitness and gradient
- Neutrality
- Neutrality Volume
- Information Content

Classification target



Target approximation

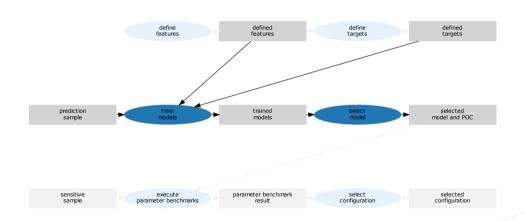
Targets

- Low end coverage *endcov*. < 0.8
- High standard deviation *stdev* > 0.1
- Relative low coverage *cov*. < *max*(*cov*.) * 0.8

Targets

target	n in class	description
<i>cov.</i> < <i>max</i> (<i>cov.</i>) * 0.8	447	End coverage less than 80% of the best execution
endcov. < 0.8	6687	End coverage less than 80%
stdev > 0.1	541	Standard deviation greater than 0.1
stdev > 0	6046	Standard deviation greater than 0.0
		The median coverage with a population
p = 125	1735	of 125 is greater than the median
		coverage with default settings
stdev > 0.1 & cov. < max(cov.) * 0.8	256	The boolean "and" of the two targets
stdev > 0 & cov. < max(cov.) * 0.8	447	The boolean "and" of the two targets
<i>stdev</i> > 0.1 & <i>endcov</i> . < 0.8	391	The boolean "and" of the two targets
pop = 125 & cov. < max(cov.) * 0.8	254	The boolean "and" of the two targets

Classification



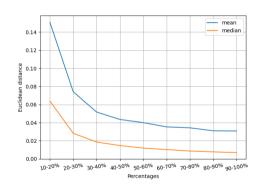
Classification

Heading

Heading

Length of Exploration Phase

- compare landscape features ever 10%
- using Euclidean distance
- trade-off between
 - time for exploration
 - time for adjusted configuration
- between 20 and 40% from search (percentage of classification)



Machine learning algorithm

Heading

- supervised learning
- fast classification
- few features
- decision tree

Heading

Hyper-parameter search

- only apply on as many as possible
 - TPR > 80%
- only apply if positive effect
 - FPR < 5%
- percentage of classification (POC)
- construction criteria (Gini Impurity, Entropy or Log-Loss)
- depth of decision tree

Component selection

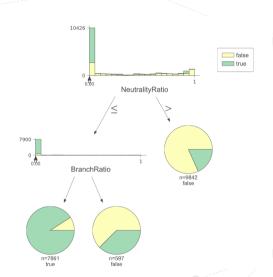
Decision tree

■ *stdev* > 0.1&*cov.max*(*cov.*) * 0.8

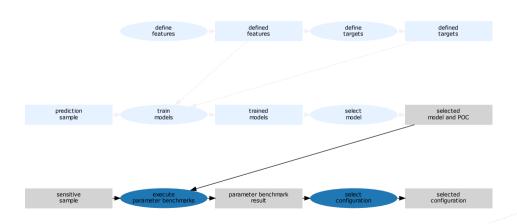
■ depth of decision tree: 2

■ Gini Impurity

■ POC: 30%



Parameter selection



Parameter selection

Heading

Heading

Component selection

			P_{adj}			\hat{P} for $k=5$				
configuration	gen.	n	mut.	cross.	ins.	mut.	cross.	ins.	<i>n</i> _+	n
default	853	9	0.95	0.68	0.05	1.0	0.87	0.0		
p = 25	1,263	13	0.96	0.69	0.04	1.0	1.0 /	0.0	0	0
p = 125	572	6	0.95	0.68	0.04	0.77	0.15	0.0	3	0
cr = 0.0	841	8	0.95	0.0	0.05	1.0	0.0	0.0	1	0
cr = 1.0	849	8	0.95	0.95	0.05	1.0	1.0 /	0.0	0	0
bias = 2.0	912	9	0.95	0.68	0.05	1.0	0.88 /	0.0	1	0
bias = 1.01	824	8	0.95	0.68	0.05	1.0	0.77	0.0	0	0
pti = 0.0	1,154	12	1.0	0.75	0.0	1.0	1.0 /	0.0	0	4
pti = 0.2	770	8	0.92	0.63	0.08	1.0	0.66	0.0	0	0
p = 10, $pti = 5.0$	912	9	0.58	0.13	0.42	0.69	0.0	0.31 /	3	0
p = 25, pti = 1.0	777	8	0.75	0.38	0.25	0.89	0.14	0.03 /	4	0

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