An Empirical Study of Incorporating Cost into Test Suite Reduction and Prioritization

[‡]Adam M. Smith and [†] Gregory M. Kapfhammer



[‡] Department of Computer Science University of Pittsburgh [†] Department of Computer Science Allegheny College



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Featuring an image from www.campusbicycle.com



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Important Contributions



Implement and empirically evaluate the efficiency and effectiveness of cost-aware greedy methods for regression test suite reduction and prioritization

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Implement and empirically **evaluate** the efficiency and effectiveness of **cost-aware** greedy methods for regression test suite **reduction** and **prioritization**

Regression Testing and Bicycles



Efficiency: Low wind resistance and time to destination

Regression Testing and Bicycles



Effectiveness: Transports all required materials and no break downs

Regression Testing and Bicycles



Cost: Frame material and components cause price to vary considerably

Regression Testing Techniques



It is **expensive** to run a test suite $T = \langle T_1, ..., T_n \rangle$. **Reduction** discards some of the *n* tests in an attempt to **decrease** testing time while still **preserving** objectives like **coverage** or **fault detection**.

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Regression Testing Techniques



It is **expensive** to run a test suite $T = \langle T_1, ..., T_n \rangle$. **Prioritization** searches through the $n! = n \times n - 1 \times ... \times 1$ orderings for those that **maximize** an objective function like **coverage** or **fault detection**.



• $R_j \rightarrow T_i$ means that requirement R_j is **covered by** test T_i

 Test suite reduction discards the test cases that redundantly cover the test requirements

• $T = \langle T_2, T_3, T_6, T_9 \rangle$ covers all of the test requirements



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ts

Γ₄ Γ₃

 T_2

 T_1

Γ₄ Γ₁ Γ₃

 T_2

Г3

 T_2

Greedy Approaches to Regression Testing



- Harrold, Gupta, Soffa (HGS)
- Delayed Greedy (DGR)
- Traditional Greedy (GRD)
- 2-Optimal Greedy (2OPT)

Hypothesis: Using the execution **time** of a test case can **improve** the reduced and prioritized test suites

Compare (i) greedy choices (cost, coverage, and ratio) and (ii) algorithms

(Regression Testing)

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Evaluating Test Suite Prioritizers



• Prioritize to **increase** the CE of a test suite $CE = \frac{Actual}{Ideal} \in [0, 1]$

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Evaluating Test Suite Reducers



Reduction Factor for Size (RFFS): How small is the reduced test suite?



Evaluating Test Suite Reducers



Reduction Factor for Time (RFFT): How fast is the reduced test suite?



	R_1	R_2	R_3	R_4	R_5	Execution Time
<i>T</i> ₁	\checkmark	\checkmark	\checkmark	\checkmark		4
<i>T</i> ₂			\checkmark	\checkmark		1
<i>T</i> ₃		\checkmark				1
T_4	\checkmark				\checkmark	1

Greedy-by	T_r	$time(T_r)$	T_p	CE
coverage	$\langle T_1, T_4 \rangle$	5	$\langle T_1, T_4, T_2, T_3 \rangle$	0.400
time	$\langle T_2, T_3, T_4 \rangle$	3	$\langle T_2, T_3, T_4, T_1 \rangle$	0.714
ratio	$\langle T_2, T_4, T_3 \rangle$	3	$\langle T_2, T_4, T_3, T_1 \rangle$	0.743

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T_4	\checkmark				\checkmark	1

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<i>T</i> ₄	\checkmark				\checkmark	1

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Case Study Applications

Name	<i>T</i>	$ \mathcal{R}(T) $	CCN	NCSS
DS	110	40	1.35	1243.00
GB	51	88	2.60	1455.00
JD	54	783	1.64	2716.00
LF	13	6	1.40	215.00
RM	13	19	2.13	569.00
SK	27	117	2.00	628.00
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Questions: Do the **greedy** reducers and prioritizers efficiently identify test suites that **improve** effectiveness? What are the fundamental **trade-offs**?



Overview of RFFT Trends





The myopic focus on cost leads to low RFFT values for 2OPT and GRD



Overview of RFFT Trends





The myopic focus on **cost** leads to **low** RFFT values for 2OPT and GRD



Overview of RFFS Trends



DGR and HGS are the best at creating test suites that improve RFFS

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Overview of CE Trends





Using ratio and cost improves the CE of the prioritized test suite

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Reduction Factor for Time - SK



For 2OPT and GRD, ratio and coverage create the best test suites



Reduction Factor for Time - SK



For 2OPT and GRD, ratio and coverage create the best test suites

Reduction Factor for Size - SK



It is often easy to construct test suites with high RFFS values

Reduction Factor for Size - SK



It is often easy to construct test suites with high RFFS values



Coverage Effectiveness Results - RP



DGR and HGS exhibit lackluster performance when reordering

Coverage Effectiveness Results - RP



DGR and HGS exhibit lackluster performance when reordering

Efficiency Measurements



For the chosen case study applications, the techniques are efficient



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For the chosen case study applications, the techniques are efficient



Alternative Evaluation Metrics Like APFD



Use **mutation** and **real** faults to support the calculation of fault detection effectiveness (**FDE**) and average percentage of faults detected (**APFD**). Consider **search-based** testing methods.

RAISE - Reduce And prlortize SuitEs

Google raise Reduce And prioritize SuitEs							Search Project	
Project Home	Downloads	Wiki	Issues	Source	Administ	er		
Summary Upd	ates							
Software developers use testing to raise confidence in the correctness of a software system. Automated reduction					🗇 Star this project			
and prioritization to				Co	de License:	Eclipse Public	License 1.0	
required to detect faults during test suite execution. This package uses the Harrold Gupta Soffa, delayed greedy, traditional greedy, and 2-optimal greedy algorithms for both test suite reduction and prioritization. Even though				Labels:		Regression, Reduction,		
						Prioritization, Software, Suite,		
						JUnit, Test, Testing		
educing and reor								
ensure that testing is cost-effective, these algorithms are				Fe	Feeds: Project Feeds			
ormally configure	d to make greedy	choices w	ith					
coverage information alone. This paper extends these				Pro	Project owners:			
Igorithms to gree					Adan	nMatthewSmith, g	kapfham	
ising both test cos code coverage to f	st (e.g. execution	ume)and in	e ratio of					

http://raise.googlecode.com/ provides tools, data sets, and resources



Concluding Remarks



- Implementation and empirical evaluation of methods for test suite reduction and prioritization
- Freely available data sets and free/open source tools

http://www.cs.allegheny.edu/~gkapfham/research/kanonizo/

