

An Investigation of Compression Techniques to Speed up Mutation Testing

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ICST 2018, April 9th, 2018

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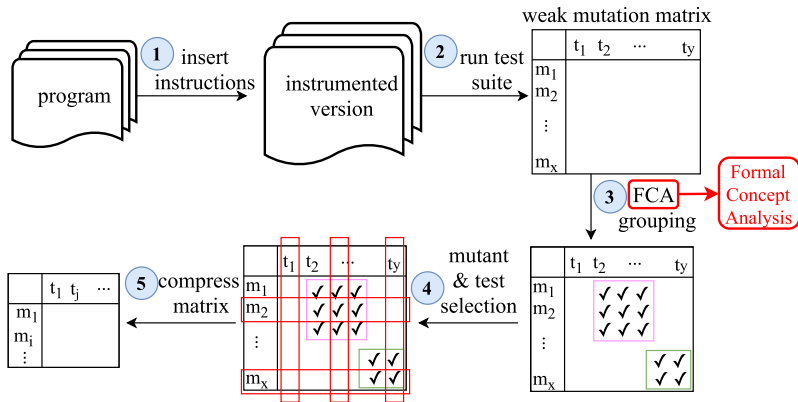
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our approach: ComMT

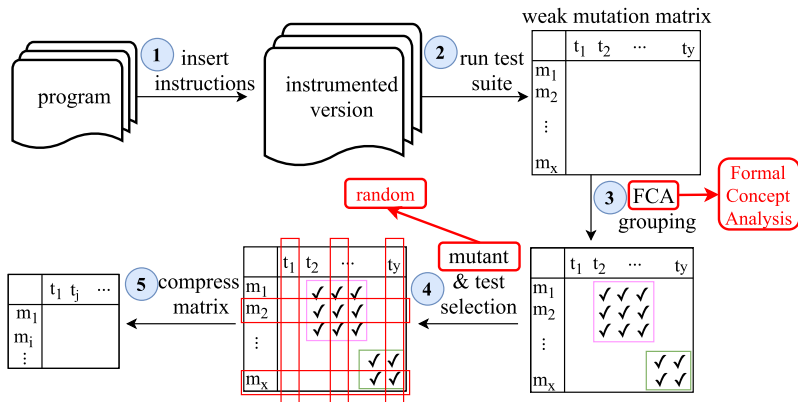
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Assumption: mutants which have high similarity in weak mutation are very likely to have the same outcome in strong mutation.



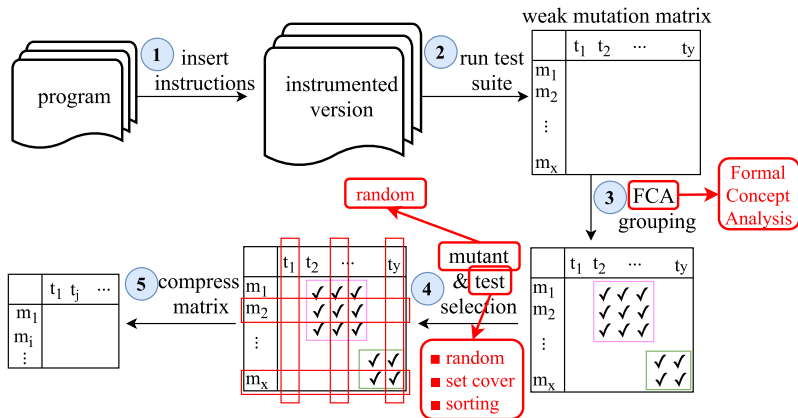
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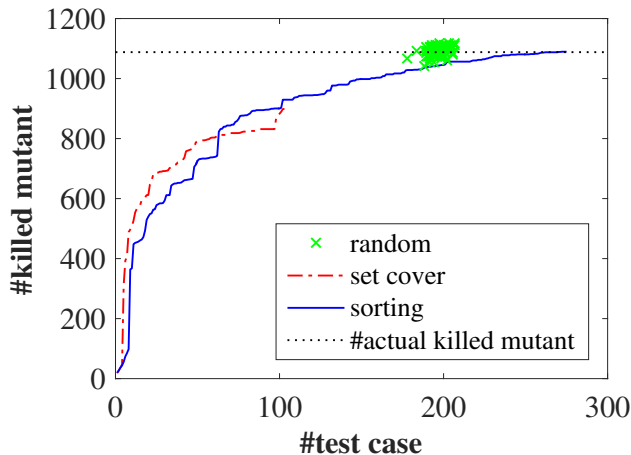


Experimental study

- **6 open-source projects**
- **automatically generated test suites (by Evosuite)**
- **comparison of test case selection**
random, set cover, sorting
- **ComMT vs. other optimisations**
coverage-based, infection-based

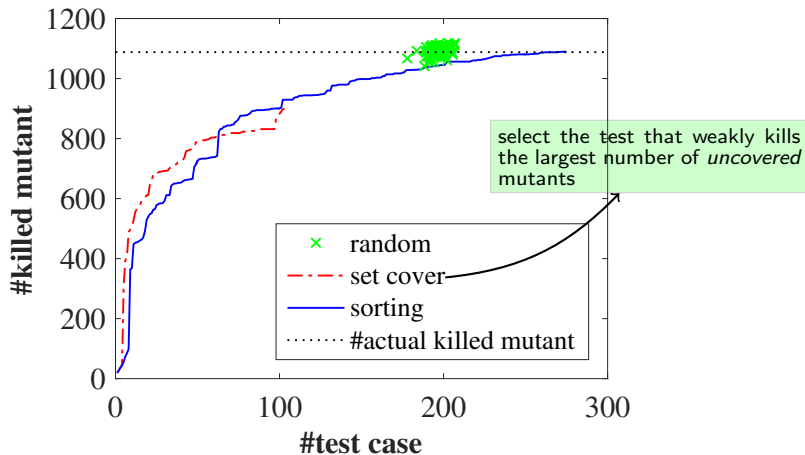
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Comparison of test case selection in each stage for jsecurity
Based on random mutant selection



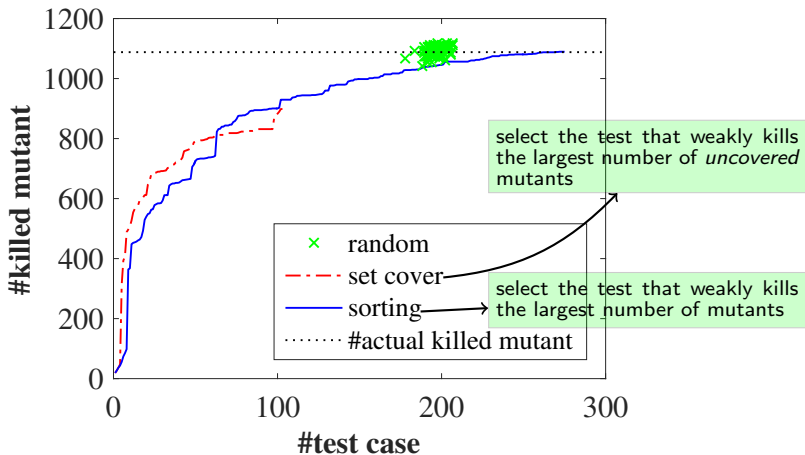
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mutant	test	red.	err.
random	no selection	83.93	0.257
	random	89.82	-19.36
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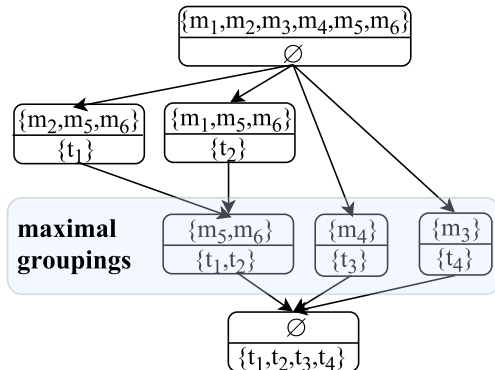
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- **Improve clustering accuracy**
E.g. explore the relationships between clustering accuracy and other metrics
- **Implement on top of an existing mutation tool**

FCA grouping

	t ₁	t ₂	t ₃	t ₄
m ₁	0	1	0	0
m ₂	1	0	0	0
m ₃	0	0	0	1
m ₄	0	0	1	0
m ₅	1	1	0	0
m ₆	1	1	0	0



Results of experimental study

- Trade-offs between execution time reduction and error rate (%):

Project	Cov. based	Inf. based	ComMT							
			No selection		Set cover		Random		Sorting	
			Red.	Err.	Red.	Err.	Red.	Err.	Red.	Err.
baseline	Exec.	Red.	Red.	Err.	Red.	Err.	Red.	Err.	Red.	Err.
jsecurity	1.39 min	16.78	87.71	0.13	90.27	-17.42	88.55	-2.46	87.31	-0.04
summa	1.54 min	13.57	90.97	0.74	93.58	-14.1	92.18	-2.25	90.87	0.69
db-everywhere	0.02 min	3.65	59.29	-0.26	80.81	-18.83	70.43	-6.22	63.95	-0.26
noen	1.58 min	6.09	88.52	0.30	91.08	-30.69	89.52	-7.29	87.98	0.14
jtaiogui	0.31 min	18.7	87.66	0.07	91.33	-13.11	89.86	-3.42	87.89	0.07
caloriecount	19.21 min	9.39	89.44	0.04	91.83	-21.99	90.5	-6.89	89.06	-0.38
Mean	-	11.37	83.93	0.257	89.82	-19.36	86.84	-4.76	84.51	0.262