Evaluating and Improving Fault Localization

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Fault localization: an important problem

Two use cases: developers and automated program repair

Many techniques and evaluations

Prior studies (winner > loser)
Ochiai > Tarantula [NLR11], [LTL], [WDGL14], [XM14], [LLT15]
Barinel > Ochiai [AZVG09]
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Op2 > Ochiai [NLR11]
Op2 > Tarantula [NLR11], [MKKY14]
DStar > Ochiai [WDGL14], [LLT15]
DStar > Tarantula [WDGL14], [JJC+14], [LLT15]
Metallaxis > Ochiai [PLT15]
MUSE > Op2 [мккү14]
MUSE > Tarantula [MKKY14]

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Prior studies (winner > loser)	Do these results hold for real
Ochiai > Tarantula [NLR11], [LTL], [WDO	world programs?
Barinel > Ochiai [AZVG09] Barinel > Tarantula (AZVG00)	
Op2 > Ochiai [NLR11]	
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Fault localization: an important problem

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Prior studies (winner > loser)	Do these results hold for real
Ochiai > Tarantula [NLR11], [LTL], [WD Barinel > Ochiai [AZVG09]	world programs? NO!
Barinel > Tarantula [AZVG09] Op2 > Ochiai [NLR11]	Why?
Op2 > Tarantula [NLR11], [MKKY14] DStar > Ochiai [WDGL14], [LLT15] DStar > Tarantula [WDGL14], [JJC+14],	 Unrealistic evaluations (artificial faults)
Metallaxis > Ochiai [PLT15] MUSE > Op2 [MKKY14] MUSE > Tarantula [MKKY14]	 Negligible or small effect sizes Unrealistic evaluation metrics

What is fault localization?



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What is fault localization?



Fault localization: how it works

Program



Spectrum-based fault localization

Program



Spectrum-based FL (SBFL)

- Compute suspiciousness per statement
- Example:

 $S(s) = \frac{failed(s)/totalfailed}{failed(s)/totalfailed + passed(s)/totalpassed}$

Statement covered by failing test Statement covered by passing test

More statement is more suspicious!

Jones et al., Visualization of test information to assist fault localization, ICSE'02

Mutation-based fault localization

Program

```
double avg(double[] nums) {
  int n = nums.length;
  double sum = 0;
  for(int i=0; i<n; ++i) {
    sum += nums[i];
  }
  return sum * n;
}</pre>
```

Mutants



Mutation-based FL (MBFL)

- Compute suspiciousness per mutant
- Aggregate results per statement
- Example:

failed(m)

 $S(s) = \max_{m \in mut(s)} \frac{f(u) e^{u(m)}}{\sqrt{totalfailed \cdot (failed(m) + passed(m))}}$

Mutant affects failing test outcome
Mutant breaks passing test

Papadakis and Traon, Metallaxis-FL: mutation-based fault localization, STVR'15

Outline and contributions

- **How to evaluate** fault localization techniques?
- **Empirical study** on artificial and real faults:
 - Do the results agree with prior work?
 - Do the results agree on artificial and real faults?
 - *No!* Explain why not.
- What design decisions matter (on real faults)?
- **How to improve** fault localization?

Evaluating fault localization techniques



Evaluating fault localization techniques



EXAM score: relative rank of the defective statement (e.g., **3/5 = 0.6**).

Smaller EXAM scores are better!



Evaluating fault localization techniques

Not straightforward for real faults:

- Multi-line defects (localize 1 or all lines?)
- Non-executable code (declarations)
- Fault of omission (>1 possible location)

Details in the paper



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Empirical study on artificial and real faults

Experimental design

- 7 widely studied FL techniques
 - **SBFL**: Barinel, D*, Ochiai, Op2, and Tarantula
 - **MBFL**: Metallaxis and Muse
- 310 real faults (5 times as many as prior studies combined)
- **2995 artificial faults** (more than prior studies combined)
- 100,000 CPU hours (MBFL is expensive)

http://www.defects4j.org

http://www.mutation-testing.org

Results of prior studies

Prior studies

	(winner > loser)
	Ochiai > Tarantula
ODEI	Barinel > Ochiai
JOFL	Barinel > Tarantula
vs. Srfi	Op2 > Ochiai
	Op2 > Tarantula
	DStar > Ochiai
	DStar > Tarantula
MBFL	Metallaxis > Ochiai
VS.	MUSE > Op2
SBFL	MUSE > Tarantula

Our results on artificial faults

	Prior studies		Ours (artificial faults)	
	(winner > l	loser)	Replicated	Effect
	Ochiai >7	Farantula	yes	small
	Barinel >	Ochiai	no	small
JDL	Barinel >7	Farantula	yes	negligible
SBEI	Op2 >	Ochiai	yes	negligible
	Op2 >7	Farantula	yes	small
	DStar >	Ochiai	yes	negligible
	DStar >7	Farantula	yes	small
MBFL	Metallaxis >	Ochiai	yes	negligible
VS.	MUSE >	Op2	no	negligible
SBFL	MUSE >7	Farantula	no	negligible

Results agree with most prior studies on artificial faults but only 3 effect sizes are not negligible.

Our results on real faults

icated Effect
nificant negligible
nificant negligible
nificant negligible
no <i>negligible</i>
nificant negligible
nificant negligible
nificant negligible
no small
no large
no large

Results disagree with all prior studies **on real faults**.

Results on artificial vs. real faults



All SBFL techniques are equally good



For SBFL, results on artificial faults **do not predict** results on real faults!

MBFL is only better than SBFL on artificial faults



For MBFL, results on artificial faults **do not predict** results on real faults!



MBFL does exceptionally well on "reversible" faults





- MBFL does exceptionally well on "reversible" faults
- Most real faults are not reversible



- MBFL does exceptionally well on "reversible" faults
- Most real faults are not reversible
- Real faults often involve unmutatable statements (e.g., break, continue, return)



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- Most real faults are not reversible
- Real faults often involve unmutatable statements

MBFL has pinpoint accuracy on artificial faults but poor performance on real faults.

Defined and explored a design space for SBFL and MBFL

• 4 design factors (e.g., formula)



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Results

• Most design decisions don't matter (in particular for SBFL)



• Definition of test-mutant interaction matters for MBFL

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Results

• Most design decisions don't matter (in particular for SBFL)



- Definition of test-mutant interaction matters for MBFL
- Barinel, D*, Ochiai, and Tarantula are indistinguishable

Existing SBFL techniques perform best. No breakthroughs in the MBFL/SBFL design space.

How to improve fault localization?



How to improve fault localization?



Explored two options:

- 1. Make MBFL great again
- 2. Hybrid: Stronger together

How to improve fault localization?



Hybrid technique is significantly better than all techniques in the MBFL/SBFL design space (small effect size).

Only top-ranked results matter

- Top-10 useful for practitioners¹.
- Top-200 useful for automated program repair².

Technique	Top-5	Тор-10	Тор-200
Hybrid	36%	45%	85%
DStar (best SBFL)	30%	39%	82%
Metallaxis (best MBFL)	29%	39%	77%
x			

Hybrid technique performs well on real use cases.

¹Kochhar et al., *Practitioners' Expectations on Automated Fault Localization*, ISSTA'16 ²Long and Rinard, *An analysis of the search spaces for generate and validate patch generation systems*, ICSE'16

Evaluating and improving fault localization

FL performance on artificial faults is not predictive for real faults.

- MBFL only better on artificial faults
- All SBFL techniques are equally good



http://bitbucket.org/rjust/fault-localization-data

http://www.defects4j.org

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MBFL/SBFL design space exploration

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- Existing SBFL techniques perform best
- No breakthroughs in the design space
 - ► FL needs to employ more information





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A new hybrid FL technique

- Combines MBFL and SBFL techniques
- Outperforms all existing FL techniques







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