

First, Fuzz the Mutants

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Abstract—The abstract goes here.

I. INTRODUCTION

A. Mutation Testing

Mutation testing [5], [3], [4] is an approach to evaluating and improving tests. Mutation testing introduces small syntactic changes into a program, under the assumption that if the original program was correct, then a program with slightly different semantics will be incorrect, and should be detected by effective tests. Mutation testing is used in software engineering research, occasionally in industry at-scale, and in some critical open-source work [1], [7], [2].

A mutation testing approach is defined by a set of mutation operators. Such operators vary widely in the literature, though a few, such as deleting a small portion of code (such as a statement), negating a conditional, or replacing arithmetic and relational operations (e.g., changing $+$ to $-$ or $==$ to $<=$), are very widely used.

II. PRELIMINARY EXPERIMENTS

Coverage differences were not statistically significant by Mann Whitney U test, but bug count differences between both methods and AFL without mutants were significant with p -value < 0.006 .

III. RELATED WORK

The most closely related work is the the T-Fuzz approach [6], which focused specifically on removing sanity checks in programs in order to fuzz more deeply. Our approach is motivated in part by the desire to remove sanity checks, but uses a more general and lightweight approach. T-Fuzz used dynamic analysis to identify sanity checks, while we simply trust that program mutants will include many (or most) sanity checks. Moreover, when a sanity check is hard to identify, but implemented by a function call, statement deletion mutants may in effect remove it where T-Fuzz will not. Our approach also introduces changes that are not within the domain of T-Fuzz, e.g., changing conditions to include one-off values. Finally, T-Fuzz worked around the fact that inputs for the modified program are not inputs for the real program under test using a symbolic execution step, while we simply hand

the inputs generated for mutants to a fuzzer and trust a good fuzzer to make use of these “hints” to find inputs for the real program, if they are close enough to be useful.

IV. CONCLUSIONS

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Method	Bugs			Stmt Cov			Branch Cov		
	Min	Max	Avg	Min	Max	Avg	Min	Max	Avg
AFL on program only	3	5	4.2	79.86%	84.37%	81.73%	78.36%	81.35%	80.40%
AFL on random mutants, non-cumulative	6	7	6.4	80.04%	84.90%	81.70%	79.85%	82.58%	80.70%
AFL on random mutants, cumulative/sequential	6	7	6.2	80.21%	84.90%	81.77%	80.10%	82.34%	80.90%
AFL on prioritized mutants, non-cumulative	6	7	6.2	81.25%	84.37%	82.39%	80.60%	81.84%	81.20%
AFL on prioritized mutants, cumulative/sequential	6	7	6.2	81.25%	84.90%	83.16%	80.10%	82.58%	81.39%

TABLE I. RESULTS FOR PRELIMINARY EXPERIMENTS