Evaluating Floating-Point Superoptimization with STOKE

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1. BACKGROUND INFORMATION

- Superoptimizers generally search through all possible programs to find the fastest version of the program supplied at the input [1]
- STOKE performs in contrast to most superoptimizers a stochastic search
- A stochastic search uses randomness to search through a subset of the entire search space
- This allows STOKE to find an optimum faster and for larger programs, however this might not be the true optimum [1]
- STOKE by default, does not formally verify the results instead it relies on randomised tests
- Floating-point errors arise normally by the order/type of operation performed because of rounding errors in between operations [2]
 - This also makes that $0.1 + 0.2 0.3 \neq 0.3 0.1 0.2$
- STOKE contains an extension that optimizes floating-point programs and allows for defining the maximum precision error [3]

2. RESEARCH QUESTION

What classes of floating-point programs cause STOKE to give well optimized results?



3. METHOD



Three small C programs



The C programs are compiled/ optimized with GCC version 4.9 and optimized with STOKE

STOKE runs multiple times with different maximum allowed errors

The C programs are compiled/optimized with GCC version 12.1

- The flag ffast-math is set this allows GCC to violate some constraints of the standard for floating point arithmetic
- The binary is also compiled without this flag



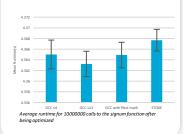
The function runtime is measured

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4. STOKE'S PERFORMANCE

Signum

- Returns the sign of the input or 0 for 0
- The resulting execution speeds where not statistically significant
- The resulting function was incorrect for specific input values
 - For instance: signum(0.0) → 1



Range Sum

double stepSize = (end - start) / n

- Sums n equally spaced values between start and end
- Can be computed without a loop By realizing that the calculation is very similar to GAUSS sum[4]
- The results would then be computed using:

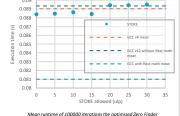
$$stepsize * \frac{n * (n + 1)}{2} + \frac{start * (start + 1)}{2}$$

- Stoke did not find this optimization
- Instead it found one that malfunctioned on specific input



Find a Zero

- Takes as input the coefficients for a **third order polynomial** $f(x) = ax^3 + bx^2 + cx + d$, the function then tries to find a **root**(an x s.t. $f(x) \approx 0$) The algorithm uses **hill climbing** in order to find an x s.t. $f(x) \approx 0$
- STOKE obtained a runtime reduction of 1.08%
 - For allowed error of 0 ulp
- Increasing allowed error resulted in slower programs



Mean runtime of 100000 iterations the optimized Zero Finde function

5. DISCUSION

- For the tests the best runtime reduction was 1.08% compared to GCC version 12.1
- During all tests STOKE was unable to generate results comparable to the original study
- Since the experiment only covers 3 algorithms are the results **Not generalizable to all** programs
 - They still can be used for answering the research question

6. CONCLUSION

- STOKE struggled to find satisfactory optimizations for all programs presented
- STOKE was never able to generate optimizations that outmatched GCC with ffastmath enabled
- The STOKE test-case generator fails to generate tests for floating-point number
 - · It failed to prevent infinite loops

7. FUTURE WORK

- Future research could focus on experimenting with different test-case generators
- To better understand the total capabilities of STOKE, future studies should focus on different program classes

 E. Schkufza, R. Sharma, and A. Aiken, "Stochastic Superoptimization," ACM SIGARCH Compute Architecture News, vol. 41, pp. 305–316, 3 2013.

Architecture News, vol. 41, pp. 305–316, 3 2013.

[2]D. Goldberg, "What Every Computer Scientist Should Know About Floating-Point Arithmetic," ACM Comput. Surv., vol. 23, no. 1, pp. 5–48, Mar 1991. doi: 10.1145/103162.103163

[3] E. Schkufza, R. Sharma, and A. Aiken, "Stochastic Optimization of Floating-point Programs With Tunable Precision," ACM SIGPLAN Notices, vol. 49, pp. 53–64, 6 2014.

[4] J. DeMaio, "Counting Triangles to Sum Squares," The College Mathematics Journal, vol. 43, pp. 297–303, Sept. 2012.Number: 4 Publisher: Taylor & Francis_eprint: https://doi.org/10.4169/college.math.j.43.4.297.