

Peppa-X: Finding Program Test Inputs to Bound Silent Data Corruption Vulnerability in HPC Applications

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Motivation: Soft Error





Silent Data Corruption (SDC)



Software Solutions



Existing Works



Heuristics-based error estimation



Our Goal

• Bound SDC probability of a program across multiple inputs

- High accuracy
- High efficiency
- In an automated way



Initial Study

- SDC probability of a program across multiple inputs
 - Vary in a large range



60%

10% 0%

- SDC probabilities of individual instructions
 - Though vary, the ranking is stable •

Pathfinder	Needle	Particlefilter	CoMD	Hpccg	Xsbench	FFT
0.92	0.79	0.90	0.90	0.96	0.59	0.77

Correlation between Rankings of Per-Instruction SDC

Prob. across Multiple Inputs

Our Approach: Overview



• Find the SDC-bound input!

Challenge #1: Deriving SDC Score

- Fuzz for small FI input
 - Small workload yet equal code coverage
 - FI simulations becomes fast
- Avoid FI simulations for all instructions
 - Reduce FI space by applying pruning
 - Use static dataflow dependency analysis
- Static dataflow dependency analysis
 - Instructions within same static data dependency shows similar SDC probabilities
 - e.g., avg FI space reduced to 49%



BB167	
<pre> %168 = load i32* %k %169 = add nsw i32 %168, 1</pre>	; ID1562; SDC: 0.8% ; ID1563; SDC: 0.4%
%171 = icmp eq i32 %169	; ID1565; SDC: 100.08
•••	

Code Example of Pruning FI space in CoMD

Pathfinder	Needle	Particlefilter	CoMD	Hpccg	Xsbench	FFT	Avg			
25.49%	51.40%	46.35%	58.44%	58.69%	49.22%	55.64%	49.32%			
FI-space pruning ratios										

Challenge #2: Fitness Function in Fuzzing

- Avoid repetitive statistical FIs to rank each generated
 - candidate input by GA
 - Assign score to each static instruction
 - Conduct FI simulations to only those instructions
 - from pruned FI space.
 - Accumulate scores of executed instructions during program

execution

• An estimate for SDC probability of a program input



Experimental Setup

- Fault N Baseline
 - Fa Generate random input to find SDC-bound input
- LLFI Inject faults to calculate SDC probability of each random input
 Rai
 - Accurate to simulate soft error and evaluate SDCs [DSN'17]
 - 100 Metrics s to evaluate SDC for each given input
 Graph Problem
 Graph Problem
 Machine Learning
 Dial
 Bigger System Solver

cation Domains

• Efficiency

Evaluation: Accuracy



- Peppa-X finds inputs that have much higher SDC probabilities than Baseline at the time budgets of selected generations
 - e.g., Xsbench: 37.9% by Peppa-X while only 0.7% by Baseline

Evaluation: Accuracy



Baseline performs as good as Peppa-X for few cases!

Evaluation: Accuracy

- The darker the color, the higher the SDC Probability
- Most colors are dark for Hpccg
 - A randomly sampled input leads to higher SDC probability
 - Easy task for Baseline!
- Most colors are light for Pathfinder
 - Difficult for Baseline to find SDC-bound inputs



space of Hpccg and Pathfinder

Evaluation: Efficiency

- What if we let baseline run for 5x more time than Peppa-X at 200 generations?
- Why to choose 200 generations?
 - Program SDC probabilities are mostly



Baseline is still unable to perform as good as Peppa-X



Program SDC probabilities bound by Peppa-X at 200 Generation and Baseline with 5x More Search Time (Y-Axis: SDC Prob., X-Axis: Benchmarks)

Use Case: Stress Test Selective Inst. Duplication

• Only a small amount of instructions being

responsible for majority of SDCs

• Duplicate only those instructions by applying 0-1

knapsack

Cost → performance overhead of an instruction if

duplicated

• Benefit \rightarrow SDC coverage by that duplicated instruction



Use Case: Stress Test Selective Inst. Duplication

• Run the protection with default reference

input and get expected SDC coverage

• Run the protected program with SDC-bound

input

Measure ac



- Inject faults w Conclusion t
 - Protection is compromised!
 - Avg expected coverage is 96.63%
 - Avg actual coverage is only 38.02%

ction level (Y-Axis: SDC Coverage, Benchmarks)

Conclusion

- Peppa-X is both accurate and efficient to identify SDC-bound inputs
 - Only one time cost for FI simulations
 - Leveraging static and dynamic analysis
- Baseline cannot find such SDC-bound inputs even with 5x more search time
 - Need extensive FI simulations to evaluate each random input
 - Not practical as FI takes long time!
- Our tool is open-sourced: <u>https://github.com/hasanur-rahman/Peppa-X</u>

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