





SAGA: A <u>Surrogate Assisted Genetic Algorithm</u> for Fast CPU Power Virus Generation

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Motivation



Both <u>manually crafting</u> and <u>automatic frameworks</u> are very time-consuming!

Where does the GA spend the time?



Can we do something better?

Surrogate Function (SF)



- Use of Surrogate Functions to accelerate power virus generation in GA-based frameworks (SAGA).
- Tested it on real-hardware across four different CPUs.
- Achieving up to 2x reduction in runtime compared to **GeST**, while achieving high quality results.

GeST [Z. Hadjilambrou, S. Das, P. N. Whatmough, D. Bull and Y. Sazeides, "GeST: An Automatic Framework For Generating CPU Stress-Tests," **2019 IEEE International Symposium on Performance Analysis of Systems and Software (ISPASS)**]

Rest of the Presentation

- Genetic Algorithm Flow
- Use of Surrogate Functions
- SAGA
- Experimental Evaluation
- Conclusion

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A collection of instruction sequences (individuals – stress tests)

Population size = 50 individuals













Genetic Algorithm's Generation Time Distribution



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The Use of Surrogate Functions (SF)



Key parameters of a Surrogate Function



Features

What input features are used to describe a virus



Prediction Model ML model used to approximate fitness



Training Set Data used to train the prediction model

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SAGA's Features

• Input features: Number of Instructions for each instruction type.



Individual (power virus):

Intuition behind SAGA's Features

• Input features: Number of Instructions for each instruction type.

- Different Instructions stress different execution ports.
- Correlate instruction mixes with power draw.
- GA discovers instruction-level parallelism to account for dependencies.



SAGA's Prediction Model – training / training set



Reference Individual



• Acts like a compass!

 It isn't a real program, but provides the characteristics of a virus with high power consumption!

SAGA's Ranking





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Platforms:

CPU	ISA	Measurement Instrument
Intel(R) Core(TM) i7-11700	x86	Likwid-powermeter
Intel(R) Core(TM) i5-2400	x86	Likwid-powermeter
Arm-Cortex-A72	Armv8-A	Arm Energy Probe
Arm-Cortex-A53	Armv8-A	Arm Energy Probe

- Integrate SAGA into GeST.
- Run each method 10 times on each platform.
- Successful threshold: 95th percentile.

SAGA vs GeST





SAGA vs GeST



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- Introduction of SF acceleration for virus generation in GA-based frameworks.
- Tested it on real-hardware across four different CPUs.
- Achieving up to 2x reduction in runtime, while achieving higher quality results.
- SAGA can be used for other types of slow optimization problems, by defining the right surrogate model.
- SAGA is publicly available and it's free!



https://www2.cs.ucy.ac.cy/carch/xi/gest_tool.php



HORIZON 2020

COMPETITIVE INDUSTRIES





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