

Abstract

- Just over a decade ago, the first review on benchmarking in Genetic Programming (GP) stated that GP needs better benchmarks [1].
- Several benchmark suites in major GP problem domains have been proposed over time.
- A diverse and accessible benchmark suite for logic synthesis is still missing.
- We take a first step towards a benchmark suite that covers different types of Boolean functions.

Logic Synthesis in Genetic Programming

- Logic synthesis (LS) in GP can be considered a black-box and optimization problem domain.
- LS by means of GP refers to the synthesis of expressions that match the input-output mapping of Boolean functions.
- LS played a major role in the application scope of GP research throughout its history.

General Motivation

- A general benchmark suite for LS is still missing in the field of GP.
- Promotion of LS to maintain a vivid application scope in GP.
- Cultivating diversity and accessibility in this problem domain.

Main Objectives and Properties

- The philosophy behind our proposed benchmark suite respects the following objectives and properties: **Generalization**, **Accessibility** and **Scalability**.
- A generalized approach to benchmarking in LS is achieved by covering a broad spectrum of types of Boolean functions.
- We make use of the scaling property of Boolean functions by proposing benchmarks of the same type with different bit lengths of the respective inputs.
- Scaling up the bit length increases the difficulty of the proposed benchmarks but maintains similarity.

Problem Selection

- We thoughtfully selected Boolean functions from seven popular categories: **arithmetic**, **transmission**, **comparison**, **counting**, **mixed**, **parity** and **cryptography**.
- We selected high dimensional multiple-output functions which remarkably differ from the overused single-output parity and multiplexer benchmarks.
- The synthesis of cryptographic Boolean functions depicts a comparatively young and interesting field of application in LS and it is therefore natural to propose these functions as benchmarks.

Interfaces and Resources

- Data files for the benchmarks are available in several formats and publicly accessible in our GitHub repository.
- Interfaces for C++, Java and Python can also be found in our repository.



Future Work

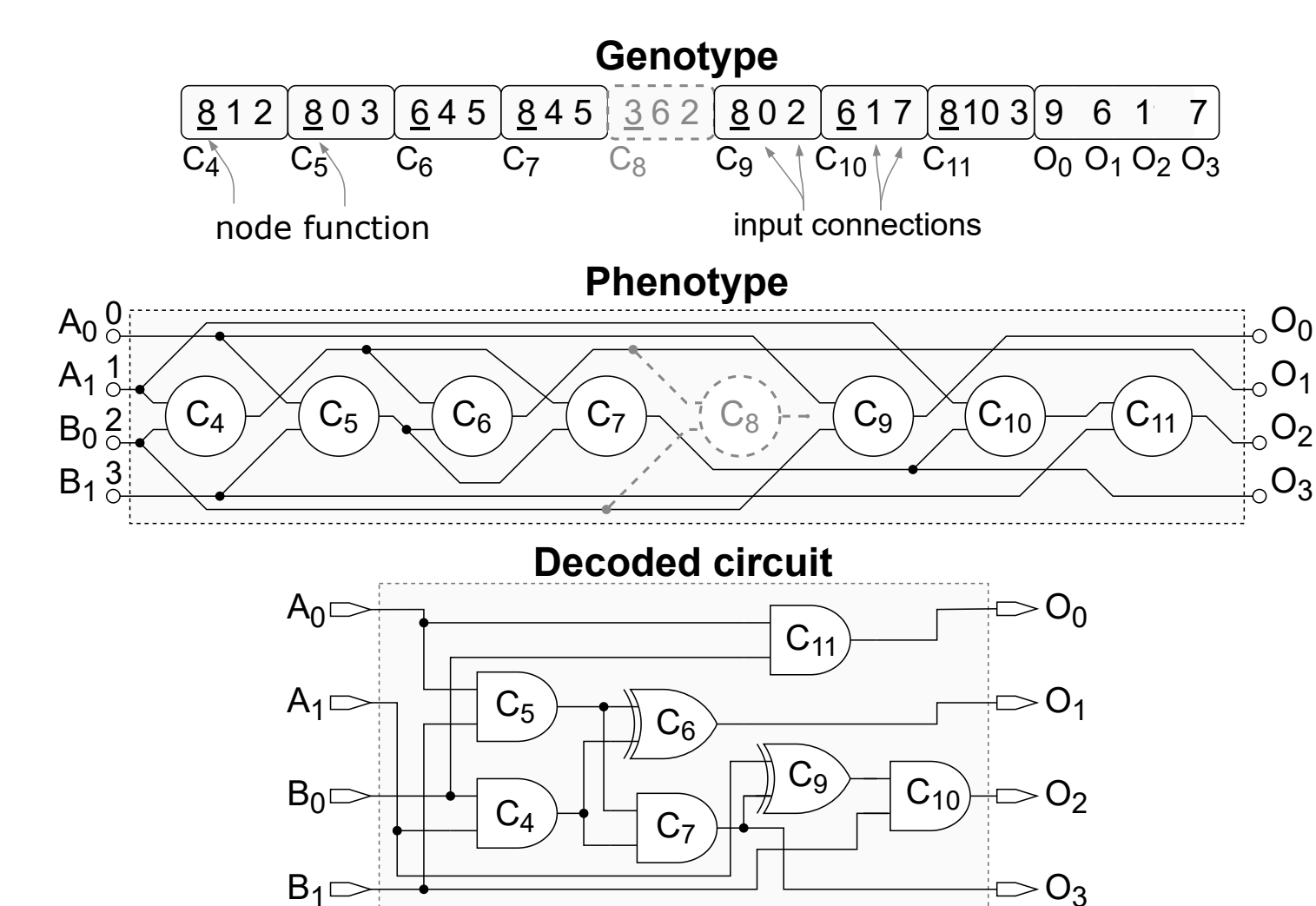
- In a **forthcoming full technical paper** we will propose our benchmark suite.
- The paper will be published in the framework of the **ACM/SIGEVO Conference on Foundations of Genetic Algorithms (FOGA)** this year.
- We also analyze the complexity of the proposed benchmarks in this paper.

List of Boolean functions that have been selected for our benchmark suite.

Arithmetic			
Problem name	Identifier	# Inputs	# Outputs
Adder 4-Bit	add4	7	4
Adder 6-Bit	add6	9	5
Adder 8-Bit	add8	11	6
Multiplier 3-Bit	mul3	6	6
Multiplier 4-Bit	mul4	8	8
Multiplier 5-Bit	mul5	10	10
Transmission			
One-hot decoder 4-Bit	dec4	4	2
One-hot decoder 8-Bit	dec8	8	3
One-hot decoder 16-Bit	dec16	16	4
One-hot decoder 8-Bit	enc8	3	8
One-hot decoder 16-Bit	enc16	4	16
One-hot decoder 32-Bit	enc32	5	32
Comparative			
Identity Comparator 5-Bit	icomp4	4	18
Identity Comparator 7-Bit	icomp5	5	30
Identity Comparator 9-Bit	icomp6	6	45
Magnitude Comparator 4-Bit	mcomp4	8	3
Magnitude Comparator 5-Bit	mcomp5	10	3
Magnitude Comparator 6-Bit	mcomp5	12	3
Counting			
Ones' counter 4-Bit	count4	4	3
Ones' counter 6-Bit	count6	6	4
Ones' counter 8-Bit	count8	8	4
Ones' counter 10-Bit	count10	10	5
Mixed (building blocks)			
Arithmetic Logic Unit 4-Bit	alu4	11	5
Arithmetic Logic Unit 6-Bit	alu6	15	7
Arithmetic Logic Unit 8-Bit	alu8	19	9
Parity			
Parity-even 8-Bit	epar8	8	1
Parity-even 9-Bit	epar9	9	1
Parity-even 10-Bit	epar10	10	1
Parity-even 11-Bit	epar11	11	1
Cryptographic			
Bent 8-Bit	Ben8	8	1
Bent 12-Bit	Ben12	12	1
Bent 16-Bit	Ben12	16	1
Balanced 8-Bit	Bal8	8	1
Balanced 12-Bit	Bal12	12	1
Resilient 8-Bit	Res8	8	1
Resilient 12-Bit	Res12	12	1
Masking 8-Bit	Mas8	8	1
Masking 12-Bit	Mas12	12	1

Baseline results

- Baseline results have been obtained with Cartesian Genetic Programming (CGP) [2].
- CGP is a graph-based representation model for GP that can be used for automated circuit design (see figure below).
- Results are available in our GitHub repository.



CGP encodes an acyclic and directed graph that can be used to represent a candidate circuit.

References

- [1] James McDermott, David R. White, Sean Luke, Luca Manzoni, Mauro Castelli, Leonardo Vanneschi, Wojciech Jaskowski, Krzysztof Krawiec, Robin Harper, Kenneth De Jong, and Una-May O'Reilly. Genetic programming needs better benchmarks. In *Proceedings of the 14th Annual Conference on Genetic and Evolutionary Computation, GECCO '12*, page 791–798, New York, NY, USA, 2012. Association for Computing Machinery.
- [2] Julian F. Miller and Peter Thomson. Cartesian genetic programming. In Riccardo Poli, Wolfgang Banzhaf, William B. Langdon, Julian F. Miller, Peter Nordin, and Terence C. Fogarty, editors, *Genetic Programming, European Conference, Edinburgh, Scotland, UK, April 15-16, 2000, Proceedings*, volume 1802 of *Lecture Notes in Computer Science*, pages 121–132. Springer, 2000.