

Genetic Synthesis: Finding the Best Specification for Synthesis Problems

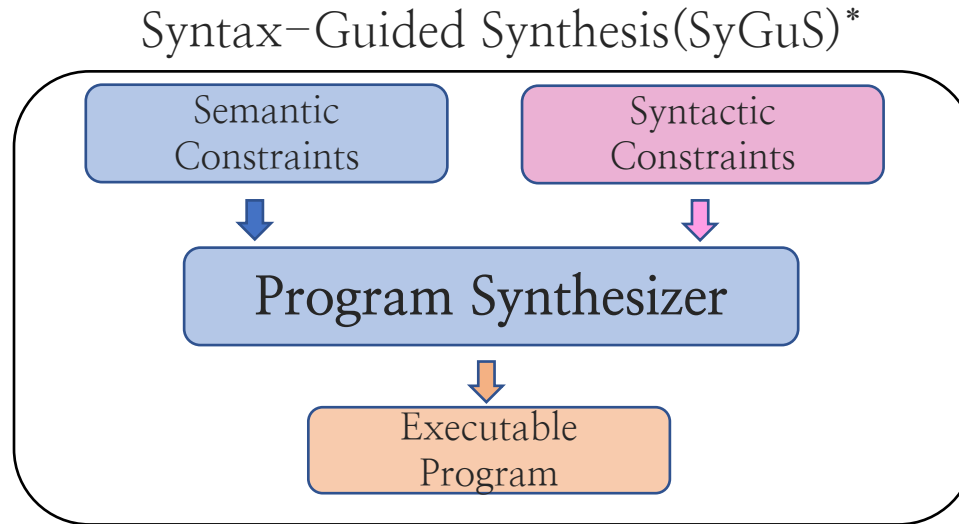
2021-2 Advanced Software Analysis
Term Project

김요엘

kimyoel2305@gmail.com

Background: What is Program Synthesis?

- The task of discovering an **executable program** from user intent expressed in the form of some **constraints**.



*Alur, Rajeev, et al. *Syntax-guided synthesis*. IEEE, 2013.

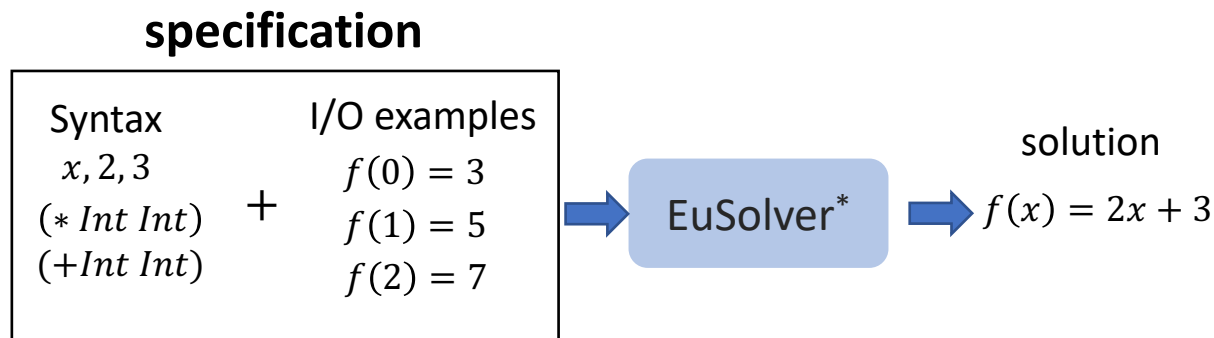
The details of Program Synthesis

The details of syntactic constraints:

- restriction of available parameters, constants, and operators (program search space).

The details of semantic constraints:

- whether candidate solutions selected in the search space satisfy given I/O examples.



*Alur, Rajeev, Arjun Radhakrishna, and Abhishek Udupa. "Scaling enumerative program synthesis via divide and conquer." *International Conference on Tools and Algorithms for the Construction and Analysis of Systems*. Springer, Berlin, Heidelberg, 2017.

The details of Program Synthesis

The details of syntactic constraints:

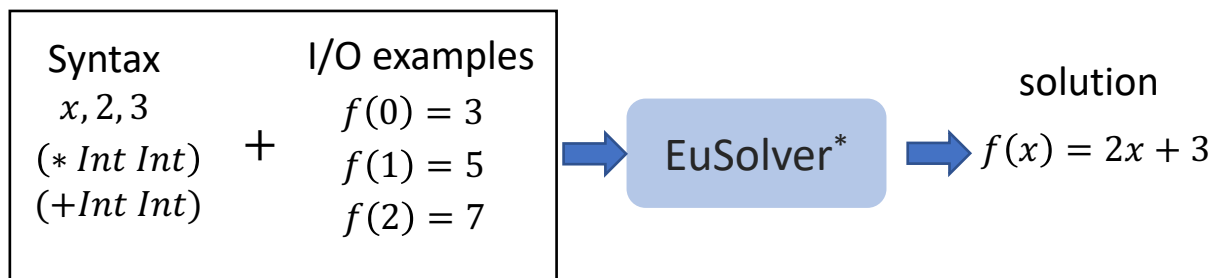
- restriction of available resources (e.g., time, money, personnel, etc.).

The details of semantic **the synthesizer to discover the best solution.**

- whether candidate scores are significantly different from each other.

My goal is to find this kind of specification that helps the synthesizer to discover the best solution.

specification



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Why is it important?

- Verification of a software containing external libraries **without source codes**.
- Static code verifiers such as symbolic execution, cannot explore these externals.
- Common alternatives: returning symbolic, non-deterministic, or random values.
- However, they could generate a lot of **false alarms** because of low accuracy.
- Note that we can get I/O examples by logging externals.
- Through program synthesis, we can synthesize externals more accurately than common alternatives.

Is it Really a Combinatorial Optimization?

- Suppose a target function f with k parameters, which uses integer/Boolean operators and constants, and returns an integer value.
- Combination of parameters: By binomial theorem, $\sum_{r=0}^k \binom{k}{r} = 2^k$.
- Combination of operators: $\{+, -, \times, \div, \%, \text{ite}, \text{or}, \text{and}, \text{not}, <, \leq, ==, >, \geq\}$, which means $\sum_{r=0}^{14} \binom{14}{r} = 2^{14}$.
- Combination of constants: $-2,147,483,648 \sim 2,147,483,647$, which means $\sum_{r=0}^{2^{32}} \binom{2^{32}}{r} = 2^{2^{32}}$.
- Theoretically, the number of syntactic constructions is $2^{k+14+2^{32}}$.

Is it Really a Combinatorial Optimization?

- We got n I/O examples after running f n times.
- Combination of I/O examples: $\sum_{r=0}^n \binom{n}{r} = 2^n$.
- In total, the number of specifications of f is $2^{n+k+14+2^{32}}$.

Design Decisions

- The size of initial population: (the number of available processors) * 2.
- Timeout for synthesizing each program (individual): 20 sec (in the experiment).
- The number of I/O examples in the initial population: 10.
- The least syntax includes '1', 'ite (if-then-else)', '==', and all parameters.
- The range of available constants in the initial population: $-1 \sim 10$.

Design Decisions

- Fitness function: the accuracy of a synthesized program. This is evaluated by executing 1,000 testcases.
- Stopping criteria: It stops if the 100% accuracy is achieved, or total execution timeout (within 15 min) is exceeded.
- Selection operator: tournament selection with $k = 5$.
- Generational selection strategy: all individuals are sorted in the order of high scores in the fitness function, and only the number of the initial population is selected as the next population.

Design Decisions (Crossover)

spec1

Syntax	I/O examples
$x, y, z, 0, 1$	$f(1,2,3) = 9$
$(+ \text{ Int Int})$	$f(2,2,4) = 11$
$(/ \text{ Int Int})$	$f(5,2,0) = 10$
$(\text{ite Bool Int Int})$	
$(== \text{ Int Int})$	

+

spec2

Syntax	I/O examples
$x, y, z, 1, 3$	$f(1,2,3) = 9$
$(* \text{ Int Int})$	$f(2,2,2) = 9$
$(\text{ite Bool Int Int})$	
$(== \text{ Int Int})$	
(or Bool Bool)	



new spec

Syntax	I/O examples
$x, y, z, 1, 3, 9$	$f(1,2,3) = 9$
$(+ \text{ Int Int})$	$f(2,2,4) = 11$
$(\text{ite Bool Int Int})$	$f(5,2,0) = 10$
$(== \text{ Int Int})$	$f(2,2,2) = 9$

	Synthesis result	Constant folding
Spec1	$x+y+z+1+1+1$	$x+y+z+3$
Spec2	$3*3$	9

Design Decisions (Mutation)

➤ adding in/output examples:

- `selected = random.choices(logs, rand(size/2));`
- `examples.addAll(selected);`

➤ deleting in/output examples:

- `selected = random.choices(examples, rand(size/2));`
- `if(size > 10) examples.removeAll(selected);`

➤ adding grammars:

- `for(operator : operators) if(25%) grammars.add(operator);`
- `if(12.5%) grammars.add(rand(100)-1);`

➤ deleting grammars:

- `for(grammar : grammars) if(12.5%) grammars.remove(grammar);`

Limitations and Future Works

➤ Limitations:

1. In some cases, it did not escape the **local maximum**.
2. The accuracy is affected by the number of I/O examples.
3. It cannot exceed the maximum performance of the synthesizer.

➤ Future works:

- We need to overcome the maximum performance of the synthesizer.
- ex) input space partitioning..