

1. Background

- Program Synthesis** is the task of finding a program based on certain specifications/constraints by the user. E.g. I/O examples, partial programs.
- Budgeted Search:** Instead of tackling synthesis as one big search problem, attack it as multiple individual searches with a limited budget, where you alter your search in some way after each attempt.
- Context Free Grammar (CFG):** Code building blocks whose combinations create the possible programs.
- Probe** is a budgeted search method where we adjust the grammar probabilities using **Just-in-Time Learning** which uses bootstrapping to train a model using an update function and partial solutions. Searching through the tree is done using **Guided Bottom-up Search** which traverses the tree using the probabilities, calculating the cost in a size-based enumeration algorithm.

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Algorithm 2 The PROBE algorithm
Input: CFG  $\mathcal{G}$ , set of input-output examples  $\mathcal{E}$ 
Output: A solution  $P$  or  $\perp$ 
1: procedure PROBE( $\mathcal{G}, \mathcal{E}$ )
2:    $\mathcal{G}_p \leftarrow \langle \mathcal{G}, p_u \rangle$            ▶ Initialize PCFG to uniform
3:    $LVL, B, E \leftarrow \emptyset, \emptyset, \emptyset$    ▶ Initialize search state
4:   while not timeout do
5:      $P, (LVL, B, E, PSol) \leftarrow \text{GUIDED-SEARCH}(\mathcal{G}_p, \mathcal{E}, (LVL, B, E, \emptyset))$  ▶ Search with current PCFG  $\mathcal{G}_p$ 
6:     if  $P \neq \perp$  then
7:       return  $P$            ▶ Solution found
8:      $PSol \leftarrow \text{SELECT}(PSol, E)$    ▶ Select promising partial solutions
9:     if  $PSol \neq \emptyset$  then
10:       $\mathcal{G}_p \leftarrow \text{UPDATE}(\mathcal{G}_p, PSol, E)$  ▶ Update the PCFG  $\mathcal{G}_p$ 
11:       $LVL, B, E \leftarrow \emptyset, \emptyset, \emptyset$  ▶ Restart the search
12:   return  $\perp$ 
  
```

[1]

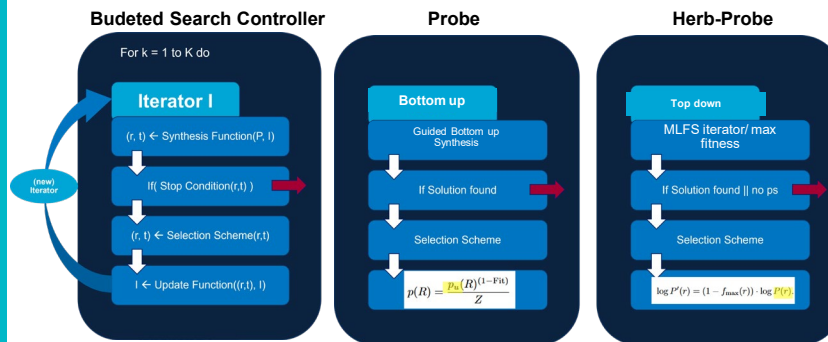
2. Research Question

How do we effectively leverage Probe-Style grammar updates, to improve program synthesis performance in Budgeted Search?

References

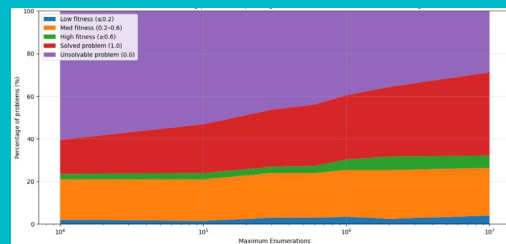
- [1] S. Barke, H. Peleg, and N. Polikarpova. Just-in-time learning for bottom-up enumerative synthesis. Proceedings of the ACM on Programming Languages, 4(OOPSLA): 1–29, Nov. 2020.
- S. Gulwani, O. Polozov, and R. Singh. Program synthesis Foundations and Trends in Programming Languages, 4(1–2):1–119, 2017.
- T. Hinnerichs, R. G. Reid, J. de Jong, B. Swinkels, P. Wochner, N. Filat, T. Magurescu, I. Hanou, and S. Dumančić. Herb.jl: A unifying program synthesis library 2025.

3. Methods



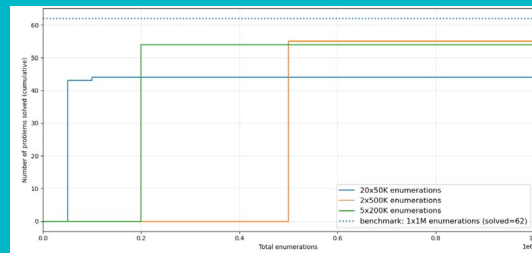
4. Evaluation

RQ1: Does increasing the total budget of a Herb-Probe run without any cycles lead to an increasing number of solved benchmarks?



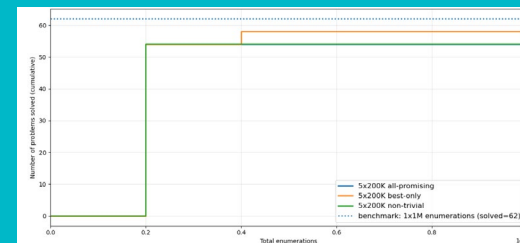
- 1x10K
- ...
- 1x1M
- ...
- 1x10M

RQ2: At a fixed total budget, do increased Herb-Probe cycles solve more benchmarks than a single search?

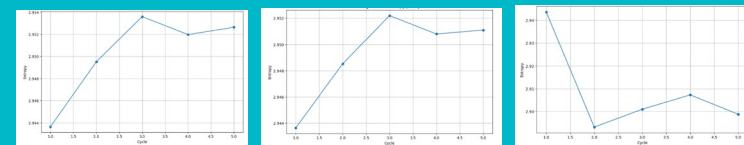


- 20x50K
- 5x200K
- 2x500K

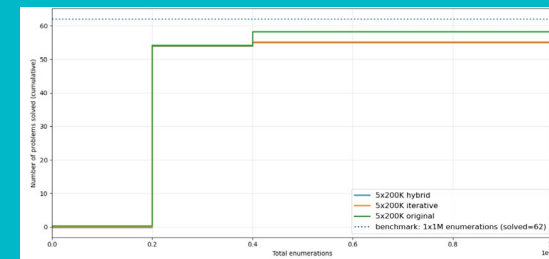
RQ3: Does the specific selector policy of Herb-Probe notably affect success?



- all promising
- non trivial > 0.2
- best only



RQ4: Does Probe's original grammar update rule remain competitive for Herb-Probe?



- Original (uniform p)
- Iterative (previous p)
- best only (normalised p)

RQ5: Does the shared framework allow comparing different approaches?

Approach	Solved Benchmarks
Herb-Probe (this work)	33 / 205
Structural Sketches	51 / 205
Promising Subprograms	31 / 205

EXTA: Benchmark vs optimal:

- 2x200K
- best_only
- original
- 1x400K
- 58
- 55

4. Conclusions & Future Work

- Strong selection pressure (best-only) is crucial
- Original Probe updater transfers best
- Currently restarts beyond cycle 2 rarely help
- Shared controller enables fair comparisons
- More drastic comparison
- Try alternative iterators, updaters (decay), more permissive selectors
- Add standard benchmarks/hybridize