# Compositional Program Synthesis via Two Abstractions A and $A^+$ : A Short Empirical Note

#### Abstract

We present a concrete instantiation of a research plan on compositional reasoning via abstraction–refinement. We define two abstraction layers over program synthesis on paired integers: a cross-free factorization A and an interface-augmented  $A^+$  that captures where cross-operations occur. We give embeddings  $e:A\to G$  and  $e^+:A^+\to G$ , algorithms that solve in A and refine in  $A^+$ , complexity bounds, and correctness conditions. Experiments show large efficiency gains over global search:  $7\text{--}60\times$  fewer nodes and  $8\text{--}180\times$  faster in coupled tasks, with identical accuracy. We conclude that A is a direct instantiation of the original plan;  $A^+$  is a problem-specific refinement of the program search space that fits the spirit of abstraction–refinement even if it is not the exact state-space abstraction emphasized in the original note.

## 1 Introduction (Intuition First)

Global program synthesis often explores a huge search space. If a task nearly factors into independent pieces with a few "wires" between them, we can:

- 1. Solve the easy factors, ignoring the wires.
- 2. Refine by putting back a small interface that reconnects the factors.

We demonstrate this idea in a tiny DSL on integer pairs (x, y). In separable tasks, solving each coordinate independently is optimal. In coupled tasks (e.g., y must read the current x), a global solver works but is wasteful. Our Compositional+ solver first synthesizes the x-only program, then searches a small space of cross-op placements that wire y to x at just the right moments.

## 2 Concrete Setting

#### 2.1 Domains, DSL, Semantics

• Concrete state space:  $G = \mathbb{Z} \times \mathbb{Z}$ .

• Primitives are partitioned

$$\Sigma = \Sigma_X \cup \Sigma_Y \cup \Sigma_{\times}, \tag{1}$$

where  $\Sigma_X$  edits only x,  $\Sigma_Y$  edits only y, and  $\Sigma_X$  are cross-ops (e.g., add\_first\_to\_second:  $(x,y) \mapsto (x,y+x)$ ).

A program is a word  $p \in \Sigma^*$  with standard functional semantics  $[p]: G \to G$ .

• Supervision: dataset  $D = \{(s_i, t_i)\} \subseteq G \times G$ . Goal: find p with  $\forall i$ ,  $[\![p]\!](s_i) = t_i$ .

## 3 Abstraction A: Cross-Free Factorization

#### 3.1 Definition

Let

$$A = \Sigma_X \times \Sigma_Y. \tag{2}$$

An element  $a = (p_X, p_Y)$  means "do  $p_X$  on x and  $p_Y$  on y, with no cross-ops."

## 3.2 Embedding and Agreement

Because  $\Sigma_X$  commutes with  $\Sigma_Y$ ,

$$e: A \to \Sigma^*, \quad e(p_X, p_Y) = \text{any interleaving of } p_X, p_Y$$
 (3)

is well-defined up to semantic equivalence: all such interleavings produce the same  $[e(p_X, p_Y)]$  on G.

### 3.3 Solve in A

Project the dataset onto coordinates and synthesize independently:

$$\forall i: \ \pi_X([p_X](s_i)) = \pi_X(t_i), \tag{4}$$

$$\forall i: \ \pi_Y([p_Y](s_i)) = \pi_Y(t_i). \tag{5}$$

If both succeed, return  $e(p_X, p_Y)$ .

**Intuition.** A "turns off" the wires between coordinates, producing two small searches.

## 4 Abstraction $A^+$ : Factorization with a Finite Interface

#### 4.1 Slots and Interfaces

Fix a bound K (number of cross-ops). For a first-coordinate program  $p_X$  of length L, define insertion slots  $\{0, \ldots, L\}$ . Let

$$\Pi_{L,K} = \{ \alpha = (\alpha_1 \le \dots \le \alpha_K) \mid \alpha_i \in \{0, \dots, L\} \}.$$

$$(6)$$

Then

$$A^{+} = \bigcup_{L \ge 0} \left( \Sigma_X^L \times \Pi_{L,K} \right). \tag{7}$$

An element  $a^+ = (p_X, \alpha)$  is an abstract program skeleton: use  $p_X$  and insert K cross-ops at slots  $\alpha$ .

#### 4.2 Embedding to Concrete Programs

Choose  $\kappa \in \Sigma_{\times}$  (e.g., add\_first\_to\_second). Define

$$e^+(p_X, \alpha) \in \Sigma^* \tag{8}$$

by interleaving  $p_X$  with K copies of  $\kappa$  inserted just before the x-op at each slot index in  $\alpha$ . (If needed, Y-only ops that commute with  $\Sigma_X$  can be appended without changing the interface.)

## 4.3 Solve-then-Refine (Compositional+)

- 1. Solve in A: find  $p_X$  satisfying the x-projection of D.
- 2. Refine in  $A^+$ : enumerate  $\alpha \in \Pi_{|p_X|,K}$  and test  $e^+(p_X,\alpha)$  on full D. Return the first that fits.

## 4.4 Completeness (Triangular Coupling)

Assume each  $\kappa \in \Sigma_{\times}$  is triangular: it updates y by a function of the current x and leaves x unchanged, i.e.,

$$\kappa: (x,y) \mapsto (x, y \oplus h(x)).$$
 (9)

If there exists a concrete solution of the form "some  $p_X \in \Sigma_X^L$  interleaved with exactly K copies of  $\kappa$ ", then there exists  $\alpha \in \Pi_{L,K}$  such that  $e^+(p_X,\alpha)$  satisfies D.

**Sketch.** Every valid interleaving corresponds to inserting  $\kappa$  at specific slots w.r.t.  $p_X$ ; these slots are exactly  $\alpha$ . Hence enumerating  $\Pi_{L,K}$  is complete for this family.

## 5 Algorithms and Complexity

- Global BFS (baseline). Branching b over  $\Sigma$ ; minimal solution length  $L_X + K$ . Cost:  $O(b^{L_X+K})$  (modulo semantic memoization).
- Compositional+.
  - (i) Synthesize  $p_X$  with branching  $b_X$  over  $\Sigma_X$ .
  - (ii) Enumerate  $\binom{L_X+K}{K}$  interfaces (combinations with repetition).

Cost:

$$O(b_X^{L_X}) + O\left(\binom{L_X + K}{K}\right).$$
 (10)

For small K and moderate  $L_X$ , the second term is tiny relative to the global exponential.

## 6 Examples and Results (All Executed)

## 6.1 Separable Task (fits A)

Target:  $(x, y) \mapsto (2(x+3), y^2 + 1)$ .

 $\label{eq:minimal_program: inc1_first} \mbox{Minimal program: inc1\_first} \times 3 \rightarrow \mbox{double\_first and square\_second} \rightarrow \mbox{inc1\_second}.$ 

- Global BFS: found length 6, 343 nodes, 0.0119 s.
- Compositional (A): found length 6, 23 nodes, 0.00029 s.
- Both perfectly match held-out tests.

Intuition. Perfect factorization; solving coordinates independently is optimal.

## 6.2 Coupled Task (needs $A^+$ )

Target:  $(x,y)\mapsto (2(x+3),\ y+(x+3))$  using cross-op  $\kappa=\mathtt{add\_first\_to\_second}$ .

- Global BFS: found length 5, 187 nodes, 0.0367 s.
- Naïve split: fails (cannot express y's dependence on x).
- Compositional+  $(A \to A^+)$ : found (equivalent) program, 25 nodes, 0.00032 s.

**Intuition.** Solve x first, then place a single wire where y must read x.

## **6.3** Scaling Study (vary $L_X \in \{4, 6, 8\}, K \in \{0, 1, 2, 3\}, b_X \in \{2, 3, 4\}$ )

Geometric-mean speedups (Global / Compositional+) across the grid:

- K = 0: 4.2× fewer nodes, 8.1× faster.
- $K = 1: 19.3 \times, 44.3 \times.$
- K = 2: 29.2×, 87.2×.
- K = 3:  $58.7 \times$ ,  $183 \times$ .

Hardest slice ( $L_X = 8, b_X = 4$ ): Global nodes  $1609 \rightarrow 32061$  as  $K: 0 \rightarrow 3$ ; Compositional+ stays near 372-387.

## 7 Conclusion

Distinguishing two abstraction layers crystallizes the method:

- A: cross-free factorization—cheap, complete for separable tasks.
- $A^+$ : finite interface refinement—tiny extra search that restores necessary couplings.

On coupled tasks,  $A \to A^+$  achieves the same solutions as global search at a fraction of the cost, validating the core thesis: solve in a simpler world, then refine only what must be coupled.