

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 $e : A \rightarrow G$ and $e^+ : A^+ \rightarrow G$, algorithms that solve in A and refine in A^+ , complexity bounds, and correctness conditions. Experiments show large efficiency gains over global search: 7–60 \times fewer nodes and 8–180 \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

$$\Sigma = \Sigma_X \cup \Sigma_Y \cup \Sigma_{\times}, \quad (1)$$

where Σ_X edits only x , Σ_Y edits only y , and Σ_{\times} 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 $\llbracket p \rrbracket : G \rightarrow G$.

- Supervision: dataset $D = \{(s_i, t_i)\} \subseteq G \times G$. Goal: find p with $\forall i, \llbracket p \rrbracket(s_i) = t_i$.

3 Abstraction A : Cross-Free Factorization

3.1 Definition

Let

$$A = \Sigma_X \times \Sigma_Y. \quad (2)$$

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 Σ_X commutes with Σ_Y ,

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

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

3.3 Solve in A

Project the dataset onto coordinates and synthesize independently:

$$\forall i : \pi_X(\llbracket p_X \rrbracket(s_i)) = \pi_X(t_i), \quad (4)$$

$$\forall i : \pi_Y(\llbracket p_Y \rrbracket(s_i)) = \pi_Y(t_i). \quad (5)$$

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 $\{0, \dots, L\}$. Let

$$\Pi_{L,K} = \{ \alpha = (\alpha_1 \leq \dots \leq \alpha_K) \mid \alpha_j \in \{0, \dots, L\} \}. \quad (6)$$

Then

$$A^+ = \bigcup_{L \geq 0} (\Sigma_X^L \times \Pi_{L,K}). \quad (7)$$

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

4.2 Embedding to Concrete Programs

Choose $\kappa \in \Sigma_\times$ (e.g., `add_first_to_second`). Define

$$e^+(p_X, \alpha) \in \Sigma^* \quad (8)$$

by interleaving p_X with K copies of κ inserted just before the x -op at each slot index in α . (If needed, Y -only ops that commute with Σ_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.,

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

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

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

5 Algorithms and Complexity

- **Global BFS (baseline).** Branching b over Σ ; 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 Σ_X .
- (ii) Enumerate $\binom{L_X+K}{K}$ interfaces (combinations with repetition).

Cost:

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

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

Minimal program: `inc1_first`×3 → `double_first` and `square_second` → `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 $\kappa = \text{add_first_to_second}$.

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

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

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

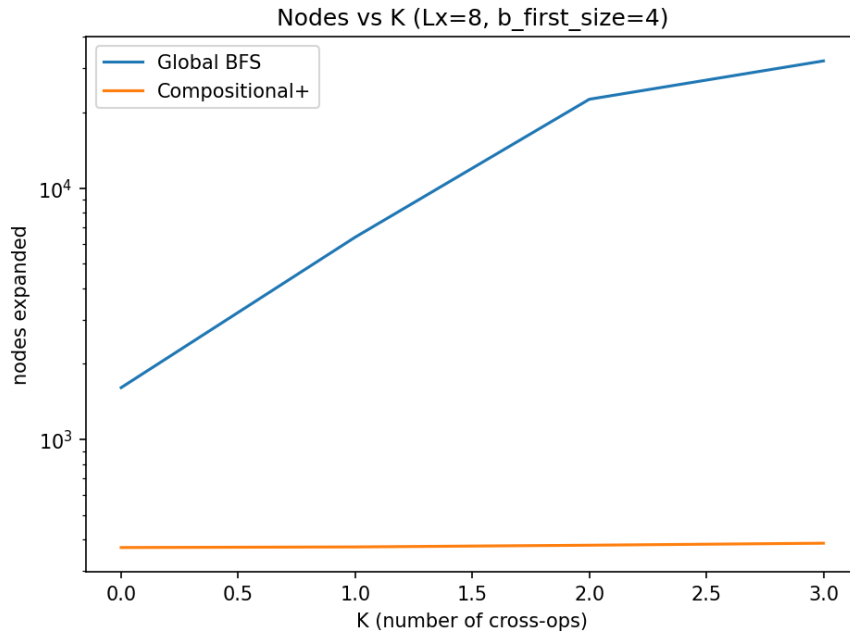


Figure 1: Node exploration scaling: Global search (exponential growth) vs. Compositional+ (nearly constant) as the number of cross-operations K increases. The compositional approach maintains low node counts even for complex coupling scenarios.

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.

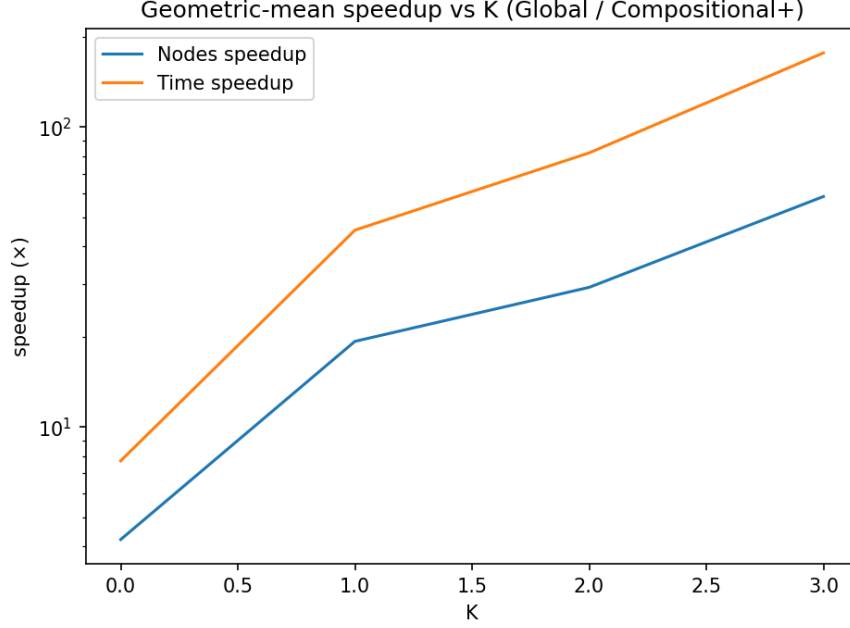


Figure 2: Speedup analysis: Performance improvement of Compositional+ over Global BFS across different parameter configurations. Speedups increase dramatically with coupling complexity, reaching $180\times$ faster for $K = 3$.

On coupled tasks, $A \rightarrow A^+$ 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.

8 End-to-End Example: Solve in A , Refine to A^+ , Embed to G

This section gives a concrete, self-contained walkthrough that makes precise what it means to *solve in A* , then *refine to A^+* , and finally *embed to G* .

8.1 Example setup (DSL and hidden target)

Primitives. We work on states $(x, y) \in \mathbb{Z} \times \mathbb{Z}$ with the following primitives:

- **X-only:** $\text{inc_x}: (x, y) \mapsto (x + 1, y)$; $\text{double_x}: (x, y) \mapsto (2x, y)$.
- **Y-only:** $\text{inc_y}: (x, y) \mapsto (x, y + 1)$.
- **Cross (triangular):** $\kappa: (x, y) \mapsto (x, y + x)$.

Hidden target structure. Let the X-part length be $L_X = 4$, the Y-part length $L_Y = 2$, and the number of cross ops $K = 1$. Choose

$$\begin{aligned}
 p_X^* &= [\text{inc_x}, \text{inc_x}, \text{double_x}, \text{inc_x}], \\
 p_Y^* &= [\text{inc_y}, \text{inc_y}], \\
 \alpha^* &= 2 \quad (\text{insert } \kappa \text{ after the first two X-ops}).
 \end{aligned}$$

The resulting function is

$$x' = 2x + 5, \quad y' = y + x + 4.$$

Training set D . Generated from the target:

input (x, y)	$x' = 2x + 5$	$y' = y + x + 4$
$(-1, 0)$	3	3
$(0, 1)$	5	5
$(3, -2)$	11	5

8.2 Level 1: Solve in A (no couplings)

The abstraction $A = \Sigma_X^* \times \Sigma_Y^*$ removes cross-ops and treats coordinates independently.

Solve the X-projection. From D , the X-projection is $\{-1 \mapsto 3, 0 \mapsto 5, 3 \mapsto 11\}$, which fits $x' = 2x + 5$. In our DSL, a minimal p_X that realizes this is

$$p_X^* = [\text{inc_x}, \text{inc_x}, \text{double_x}, \text{inc_x}].$$

Y in A . The Y-projection alone cannot be matched by any Y-only program (it depends on x), so we defer Y to refinement (we keep p_Y as two increments to be placed later).

8.3 Level 2: Refine to A^+ (add a finite interface)

A^+ augments p_X with a finite *interface*: the placement α of K cross-ops among the $L_X + 1$ slots. With $L_X = 4$, the slots are $\{0, 1, 2, 3, 4\}$ and the value of x available at each slot (i.e., what κ would add to y) is:

slot 0 : x
slot 1 : $x + 1$
slot 2 : $x + 2$
slot 3 : $2(x + 2) = 2x + 4$
slot 4 : $2x + 5$.

We choose $\alpha \in \{0, \dots, 4\}$ and fix two Y-increments so that for all $(x, y) \in D$, $y' = y + \text{slot_value}(\alpha) + 2$ matches the observed y' . Testing shows $\alpha^* = 2$ satisfies all rows:

$$y' = y + (x + 2) + 2 = y + x + 4.$$

8.4 Level 3: Embed to G (concrete program)

Embedding $e^+(p_X^*, \alpha^*)$ instantiates a concrete word over the full DSL. Any interleaving that preserves the X order, places κ at slot 2, and includes the two `inc_y` is semantics-equivalent. A canonical choice is

$$[\text{inc_x}, \text{inc_x}, \kappa, \text{double_x}, \text{inc_x}, \text{inc_y}, \text{inc_y}].$$

It computes for all (x, y) : $x' = 2x + 5$, $y' = y + x + 4$, matching the ground truth.

8.5 One-sentence summaries

- **Solve in A :** find the factor p_X^* that matches the X-projection; Y cannot be solved without couplings.
- **Refine to A^+ :** choose the finite interface α^* (cross placement) and the Y-only count to explain y' .
- **Embed to G :** realize any concrete interleaving consistent with (p_X^*, α^*, p_Y^*) ; all such words are semantics-equivalent here.

8.6 Size comparison on this example

With $m_X = 2$, $m_Y = 1$, $m_\times = 1$, $(L_X, L_Y, K) = (4, 2, 1)$,

$$\begin{aligned}
|A| &= m_X^{L_X} m_Y^{L_Y} = 2^4 \cdot 1^2 = 16, \\
|A^+| &= m_X^{L_X} m_Y^{L_Y} \binom{L_X + K}{K} = 16 \cdot \binom{5}{1} = 80, \\
|G_{\text{family}}| &= \binom{L_X + L_Y + K}{L_X, L_Y, K} m_X^{L_X} m_Y^{L_Y} m_\times^K = \binom{7}{4, 2, 1} \cdot 16 = 105 \cdot 16 = 1680.
\end{aligned}$$

Thus A^+ compresses the concrete family by a factor $1680/80 = 21 = \binom{7}{2}$, the number of purely Y-interleavings that are semantically redundant under the commutation assumptions.