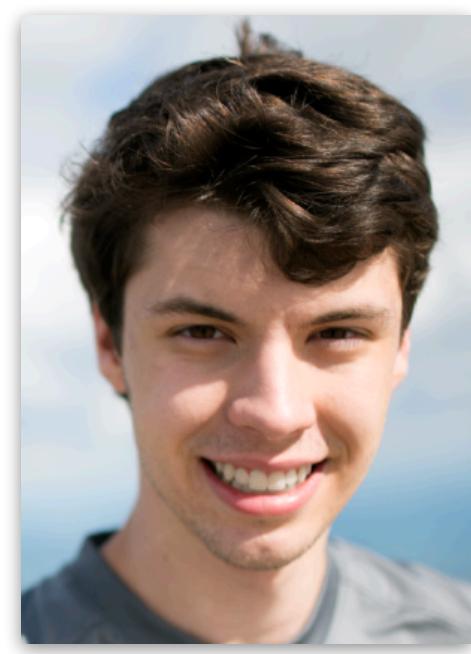


# Poseidon: Automatic Profile-Guided Optimizations of Floating-Point Programs in Compilers



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# Numerical Analysis Primer

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- A trend in Machine Learning: reduce precision to increase performance!



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- A trend in Machine Learning: reduce precision to increase performance!
- Precision changes are not always the answer.
- Example:  $1.0 \times 10^8 + 1.0 - 1.0 \times 10^8 = ?$
- FP32:  $(1.0 \times 10^8 + 1.0) - 1.0 \times 10^8 = 1.0 \times 10^8 - 1.0 \times 10^8 = 0.0$



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  - FP32:  $(1.0 \times 10^8 + 1.0) - 1.0 \times 10^8 = 1.0 \times 10^8 - 1.0 \times 10^8 = 0.0$
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  - FP64:  $(1.0 \times 10^8 + 1.0) - 1.0 \times 10^8 = 1.00000001 \times 10^8 - 1.0 \times 10^8 = 1.0$
  - **FP32 Reassociated:**  
 $(1.0 \times 10^8 + 1.0) - 1.0 \times 10^8 = (1.0 \times 10^8 - 1.0 \times 10^8) + 1.0 = 1.0$



# Existing Approaches

---

- Precision Tuning
  - $F32 \rightarrow BF16, F16, F64, F128, \dots$
- Floating-Point Expression Rewrites
  - $\sqrt{x+1} - \sqrt{x} \rightarrow \frac{1}{\sqrt{x+1} + \sqrt{x}}$
- Math Library Functions
  - $\sin(x) \approx x - \frac{x^3}{3!} + \frac{x^5}{5!}$



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All require human intervention to determine if they are valid, let alone performant!



Can we automate existing FP optimization techniques in compilers?

How much performance are we sacrificing due to *poor* choices of  
FP precision and expressions?



# Poseidon

- An end-to-end system which automatically transforms floating-point expressions in programs to maximize program accuracy subject to a computation cost budget
- Implemented as a PGO-like two-phase compilation within LLVM
  - First Compilation: **Instrumentation Pass**
  - Second Compilation
    - **FP Subgraph & Expression Transformations**
    - **Dynamic Programming Solver**



Floats...



# Poseidon

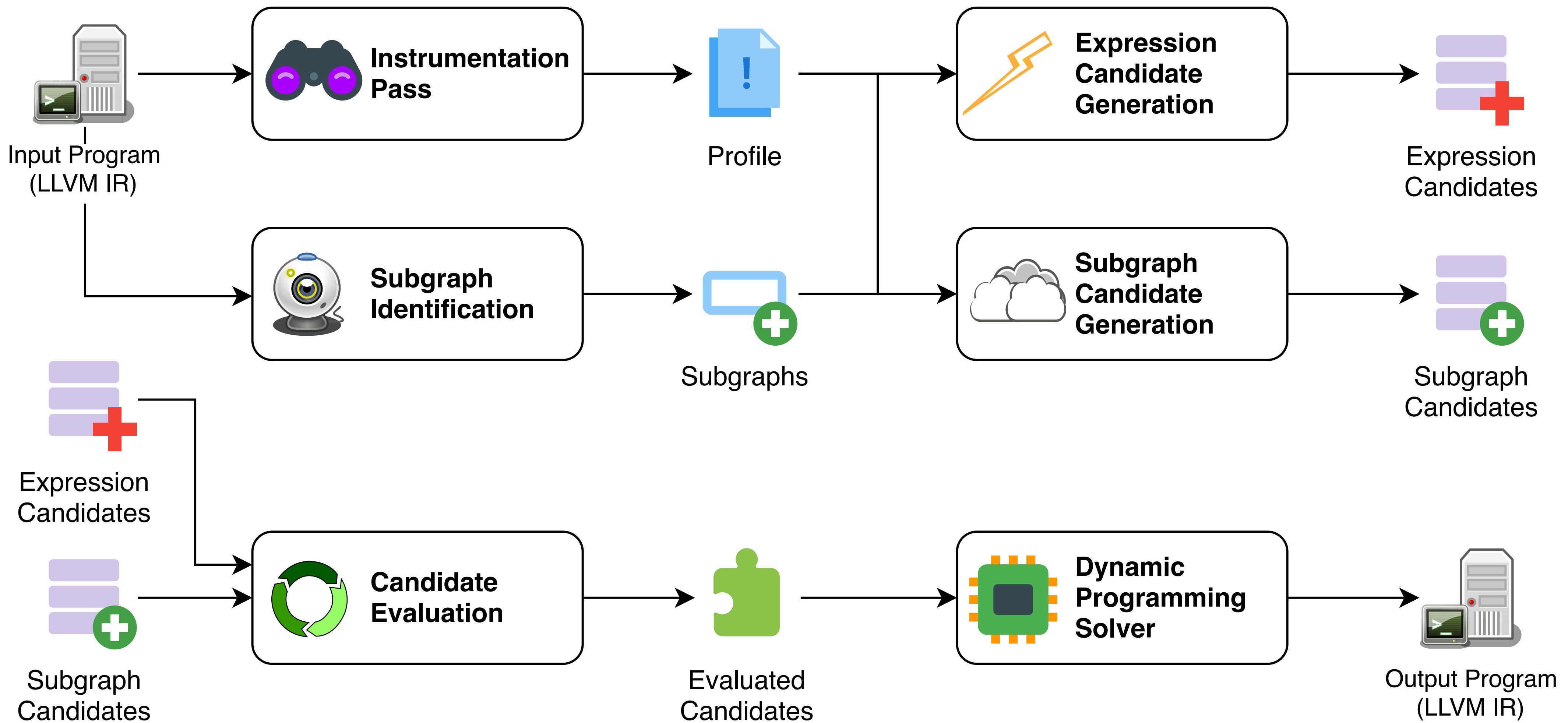
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- Implemented on top of **Enzyme**, a High-Performance Automatic Differentiator of LLVM and MLIR



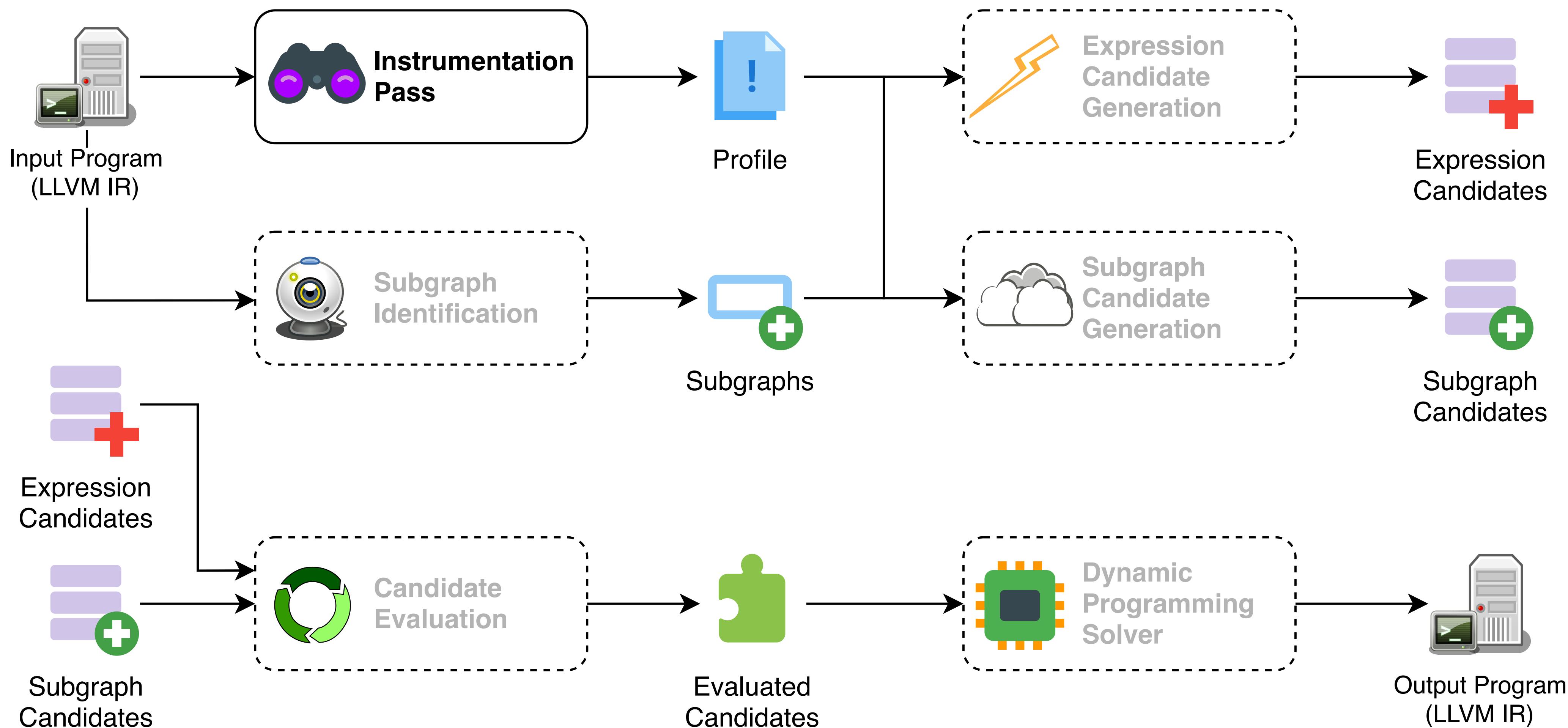
Floats...



# Poseidon Overview



# Phase 1: Instrumentation



# Phase 1: Instrumentation

---

- Augment the input program
  - Logger function calls embedded in Enzyme-synthesized gradient
- Run the augmented program to extract information of FP instructions



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- Augment the input program
  - Logger function calls embedded in Enzyme-synthesized gradient
- Run the augmented program to extract information of FP instructions

```
double example(...) {  
    double tmp;  
    if (a < b)  
        tmp = sin(a + c * 3.0);  
    else  
        tmp = cos(b + c * 4.0);  
    return tmp;  
}
```

```
define double @diffe_example(...) {  
    ...  
    %tmp = call @llvm.sin.f64(...)  
    call void @enzymeLogValue(double %tmp, ...)  
    %grad = call @llvm.cos.f64(...)  
    call void @enzymeLogGrad(double %grad, ...)  
    ...  
}
```

Original computation



# Phase 1: Instrumentation

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Enzyme-synthesized gradient



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    ...  
}
```

Enzyme-synthesized gradient

Logger function calls



# Phase 1: Instrumentation

- Augment the input program
  - Logger function calls embedded in Enzyme-synthesized gradient
- Run the augmented program to extract information of FP instructions

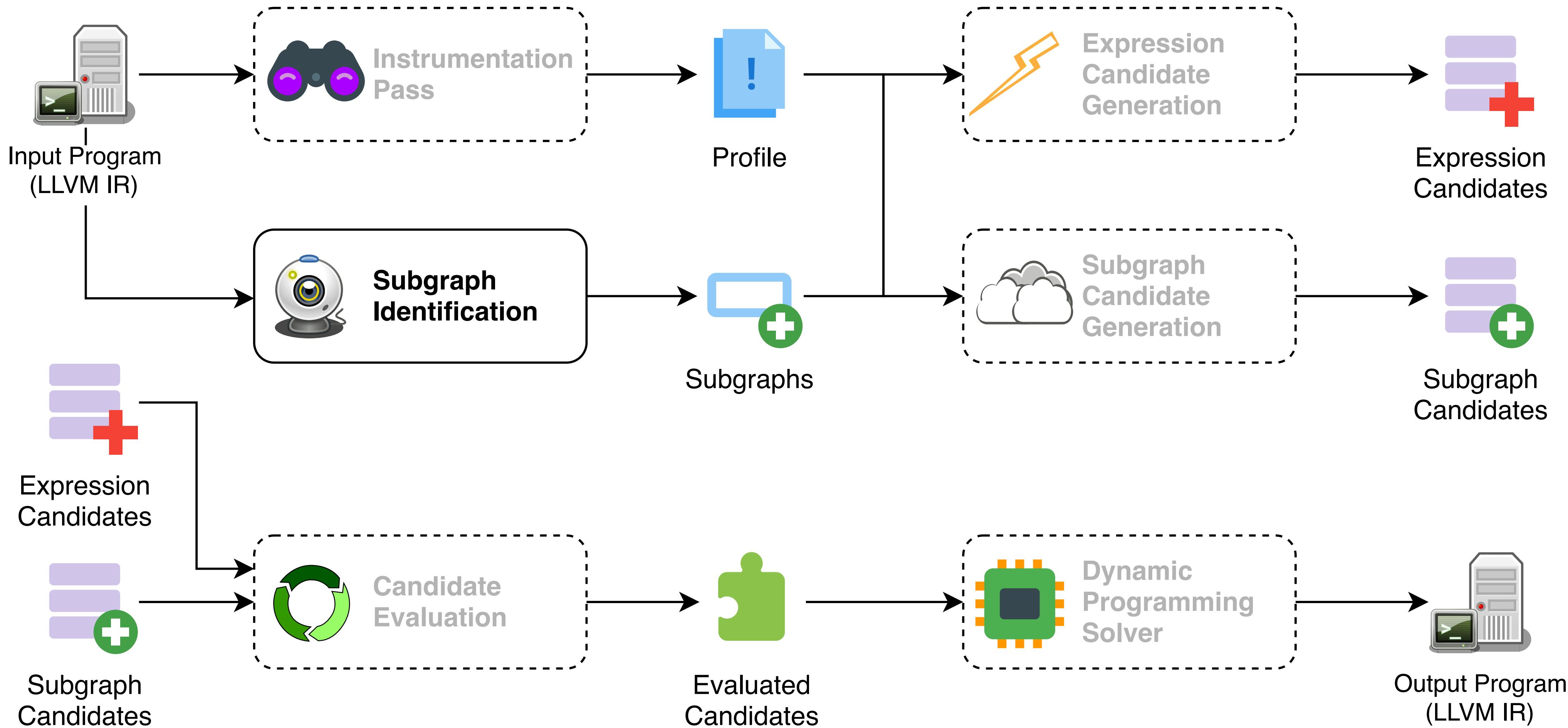
```
double example(...) {  
    double tmp;  
    if (a < b)  
        tmp = sin(a + c * 3.0);  
    else  
        tmp = cos(b + c * 4.0);  
    return tmp;  
}
```

```
Value:example:1:2  
    Min = -0.999...  
    Max = 0.999...  
    Executions = 50053  
    Geometric Average = 6.017...e-01  
// Other Values...  
Grad:example:1:2  
    Geometric Average = 9.999...e-01  
// Other Gradients...
```

Original computation



# Phase 2A: Subgraph Identification



# Phase 2A: Subgraph Identification

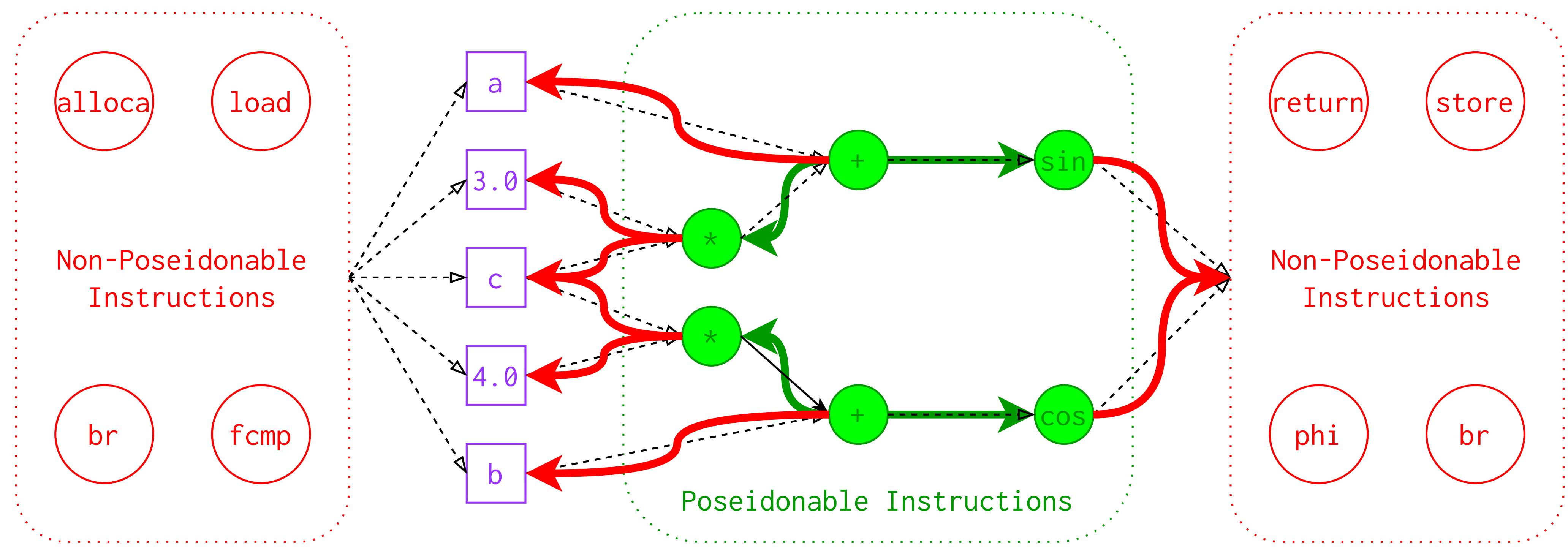
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- Optimizable `llvm::Value` → *Poseidon-able* Values
  - Poseidonable Values (FP Instructions)
    - `+, -, *, /, sin, cos, tan, exp, log, sqrt, cbrt, pow, fma, hypot, expm1, log1p, ...`
  - Non-Poseidonable Values
    - `load, store, br, ret, if*, ==*, <*, >*, <=*, >=*, !=*, and*, or*, not*, ...`
- \* Opcode planned but not yet analyzable in Poseidon.



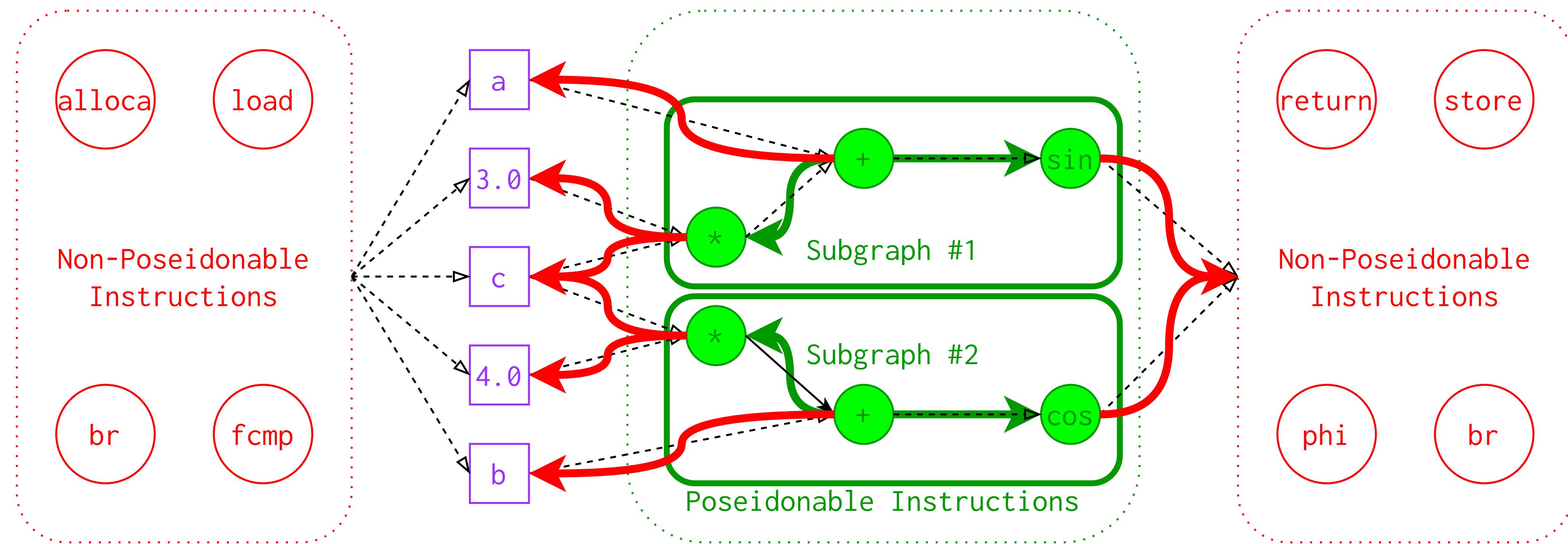
# Phase 2A: Subgraph Identification

- Floodfill Algorithm → DAGs of Poseidonable Instructions

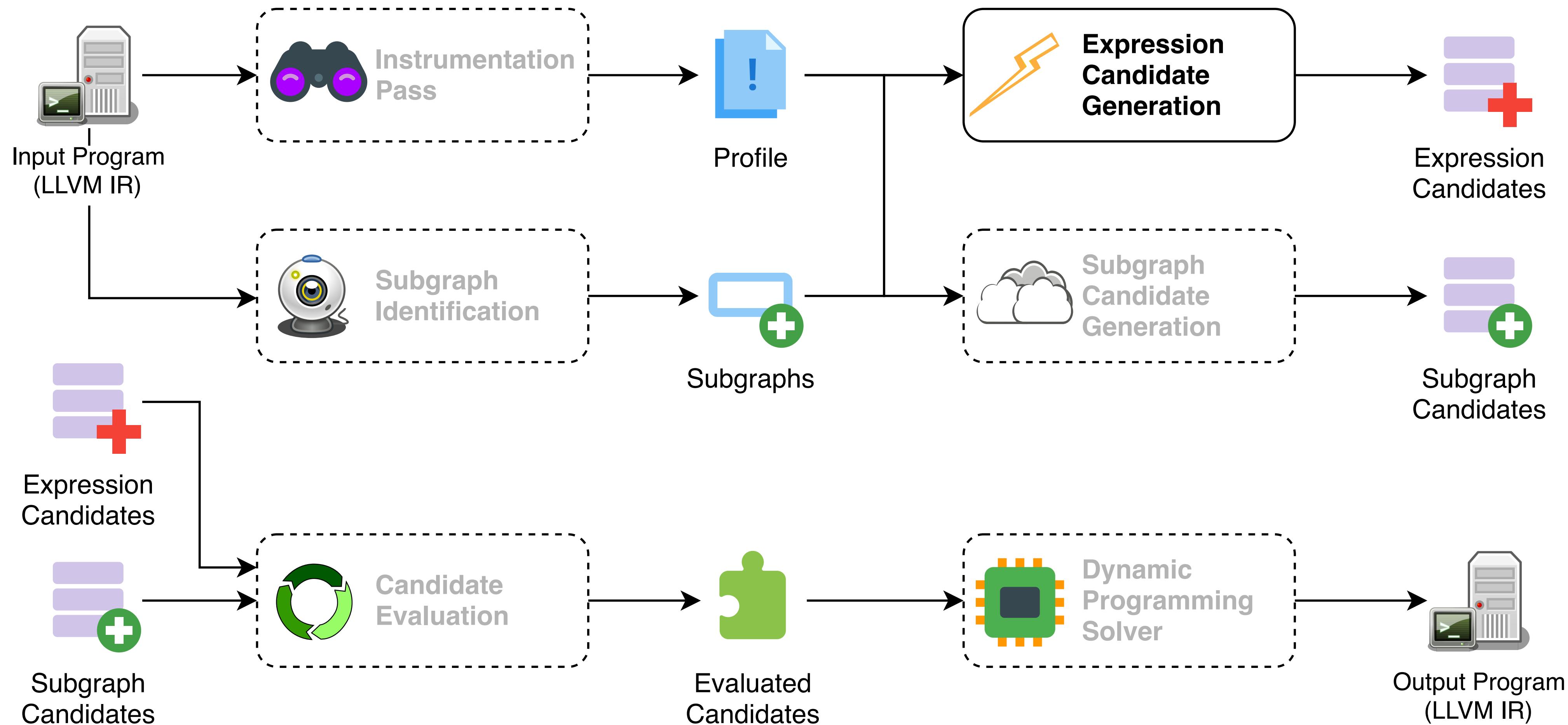


# Phase 2A: Subgraph Identification

- Floodfill Algorithm → DAGs of Poseidonable Instructions

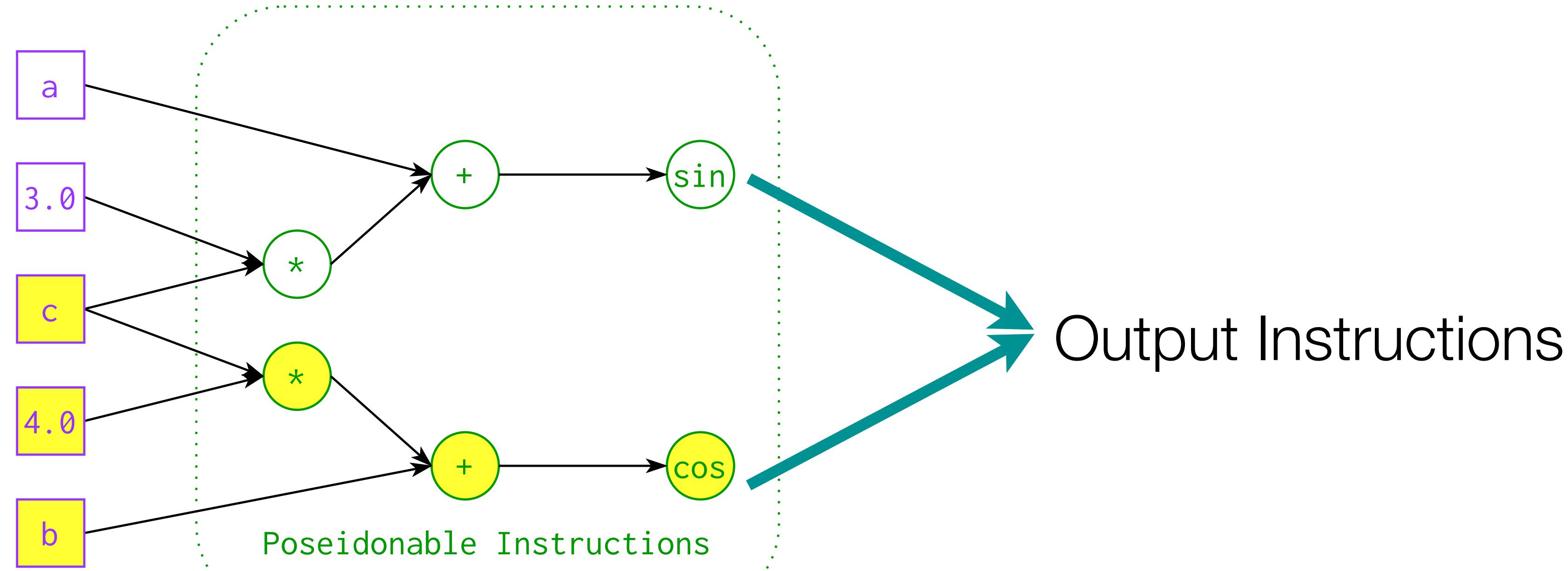


# Phase 2B: Expression Candidate Generation



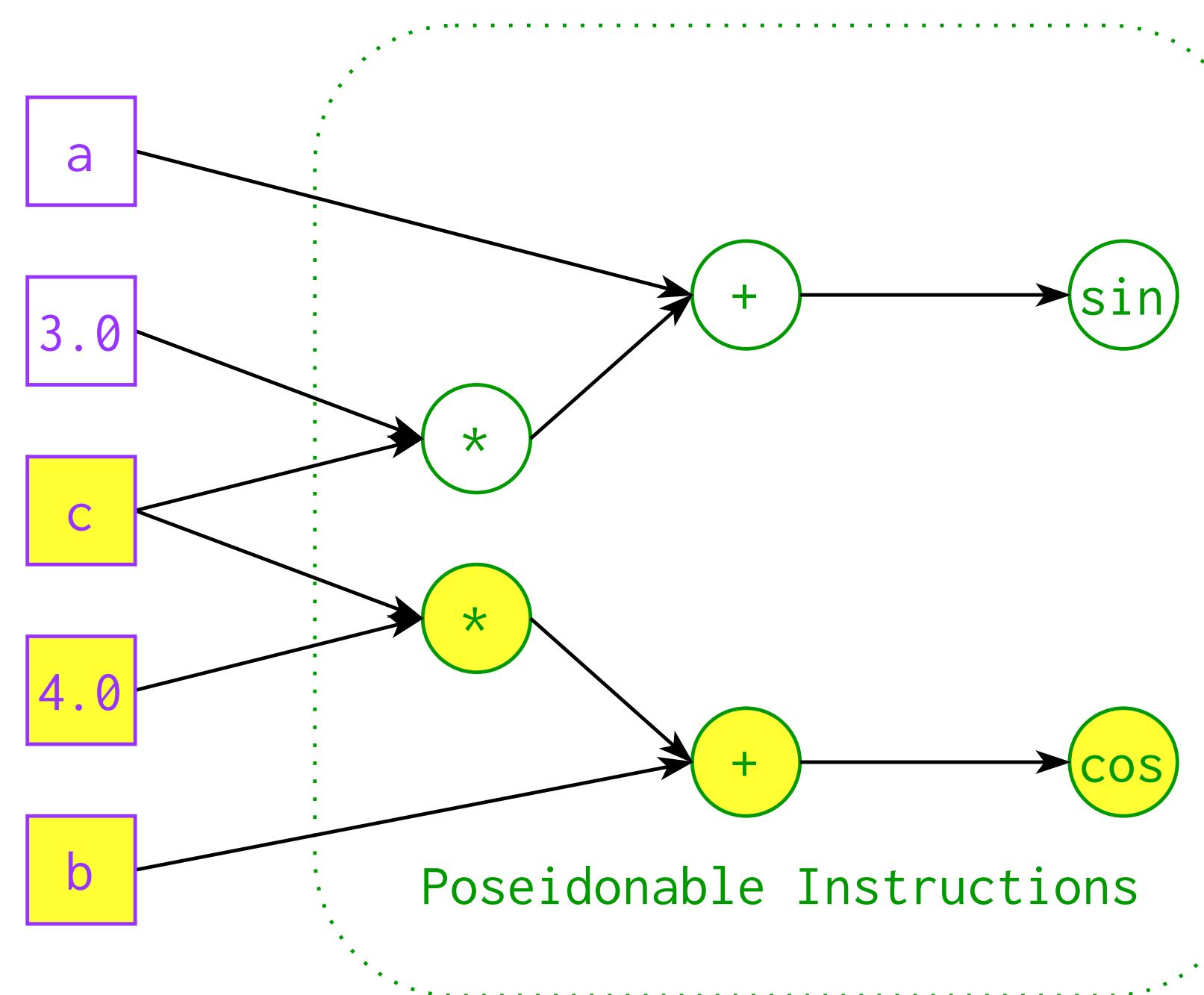
# Phase 2B: Expression Candidate Generation

- *Output instructions* propagate values to non-poseidonable values



# Phase 2B: Expression Candidate Generation

- *Output instructions* propagate values to non-poseidonable values
- Construct  $\mathbb{F}$  expressions from *output instructions* and use existing tools to produce expression candidates



```
// Herbie* Input
(FPCore (v0 v2)
:pre (and (<= 1.0...e+00 v0 8.9...e+00)
           (<= 1.0...e+00 v2 8.9...e+00))
      (cos (+ (* v0 4) v2)))
```

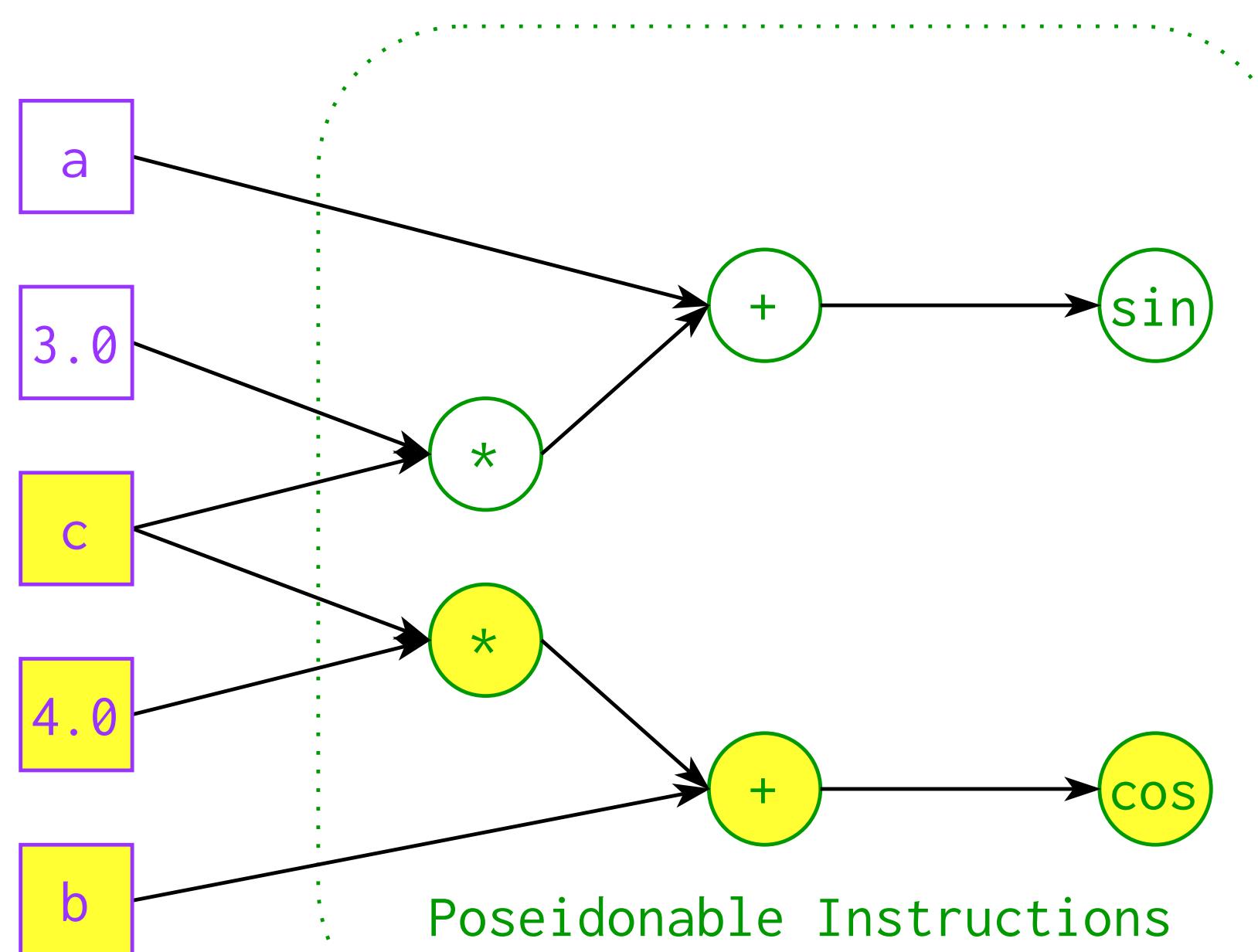
Bounds from Phase 1: Instrumentation.

Constructed expression



# Phase 2B: Expression Candidate Generation

- *Output instructions* propagate values to non-poseidonable values
- Construct  $\mathbb{F}$  expressions from *output instructions* and use existing tools to produce expression candidates

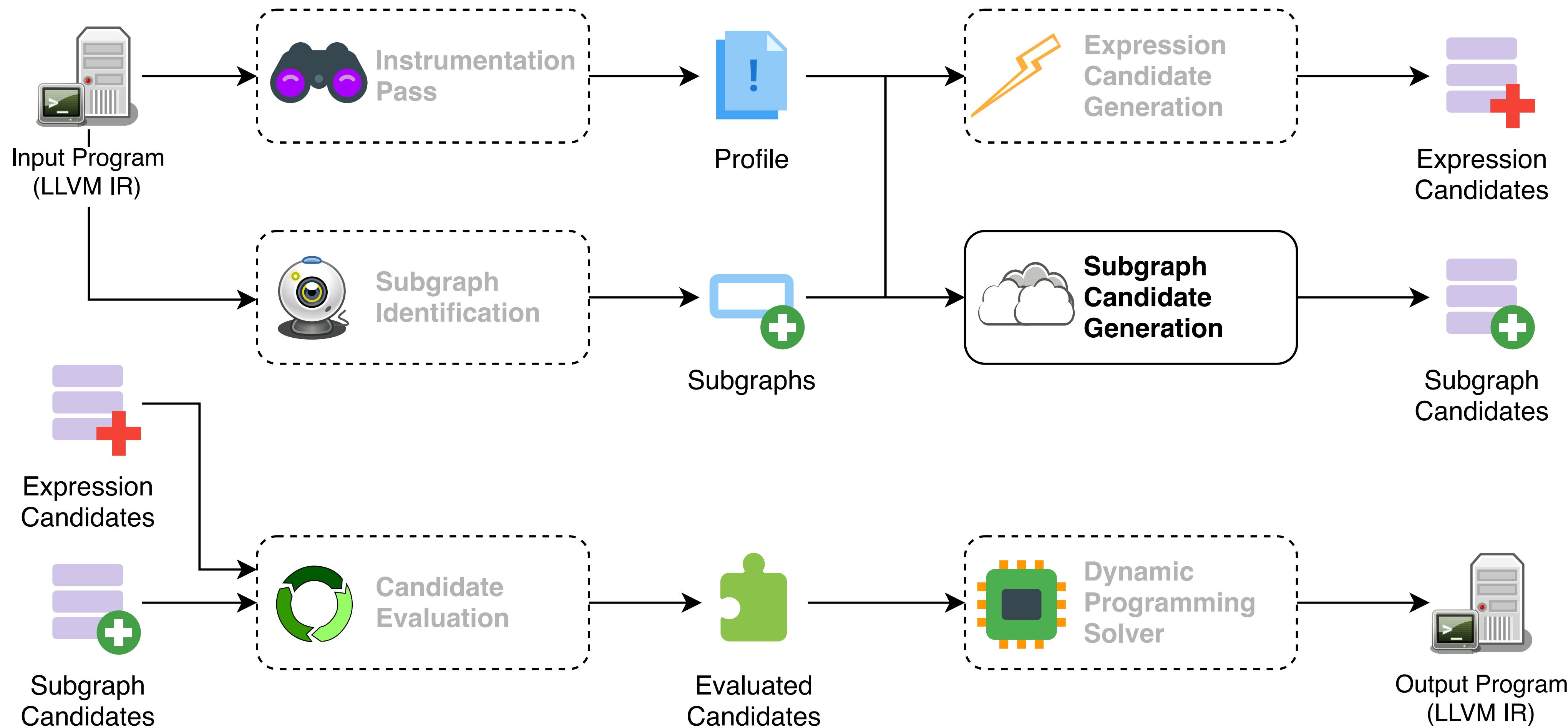


```
// Herbie* Input
(FPCore (v0 v2)
:pre (and (<= 1.0...e+00 v0 8.9...e+00)
           (<= 1.0...e+00 v2 8.9...e+00))
      (cos (+ (* v0 4) v2)))
```

```
// Expression Candidate #1
(- (* (cos (* v0 4)) (cos v2))
   (* (sin (* v0 4)) (sin v2)))
// Expression Candidate #2
(fma (cos (* v0 4)) (cos v2) (*
   (sin (* v0 4)) (sin (neg v2))))
```



# Phase 2C: Subgraph Candidate Generation



# Phase 2C: Subgraph Candidate Generation

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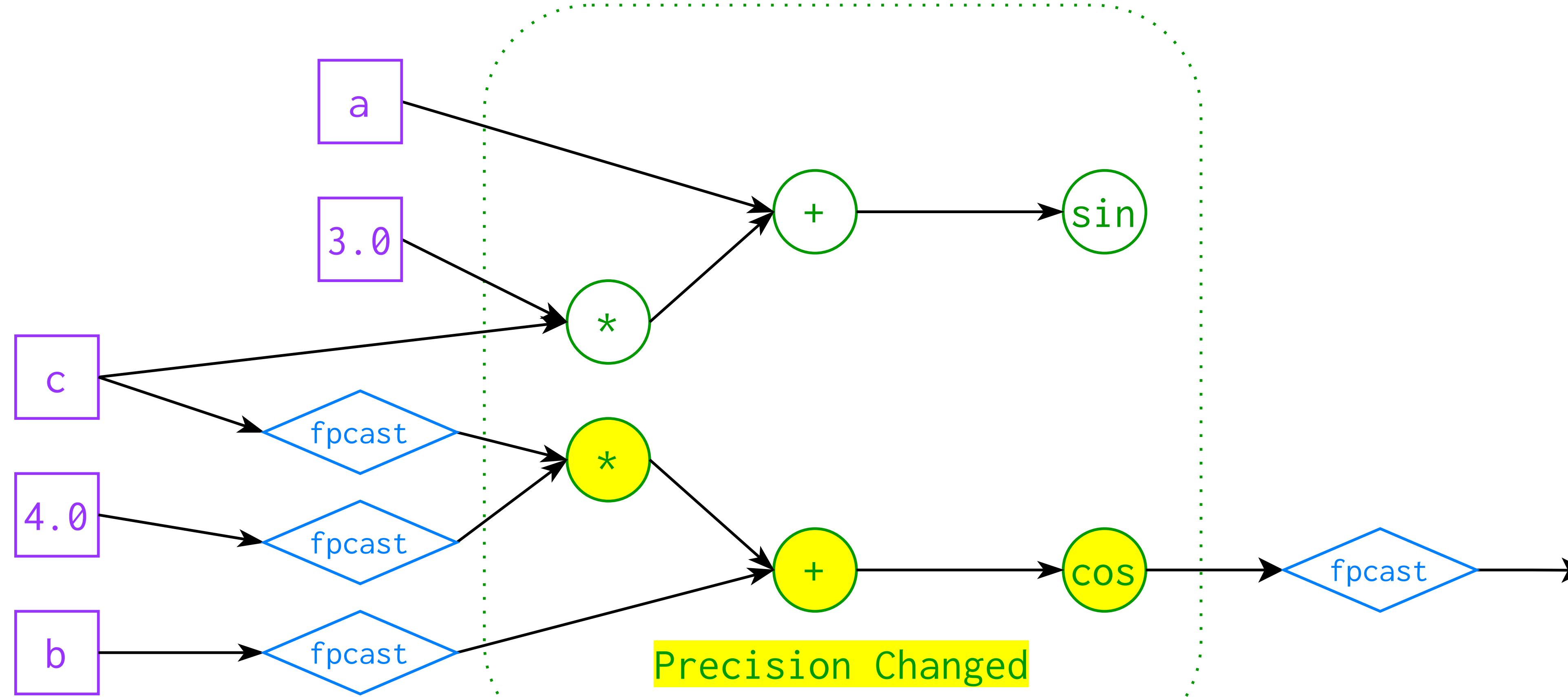
- **Sensitivity\*** (How much changing an intermediate instruction impacts the final result)
  - Computed as  $|x \cdot f'(x)|$
  - Values and gradients come from *Phase 1: Instrumentation*
- Gradient-Guided Precision Tuning

\* From ADAPT (Menon et al. '18).

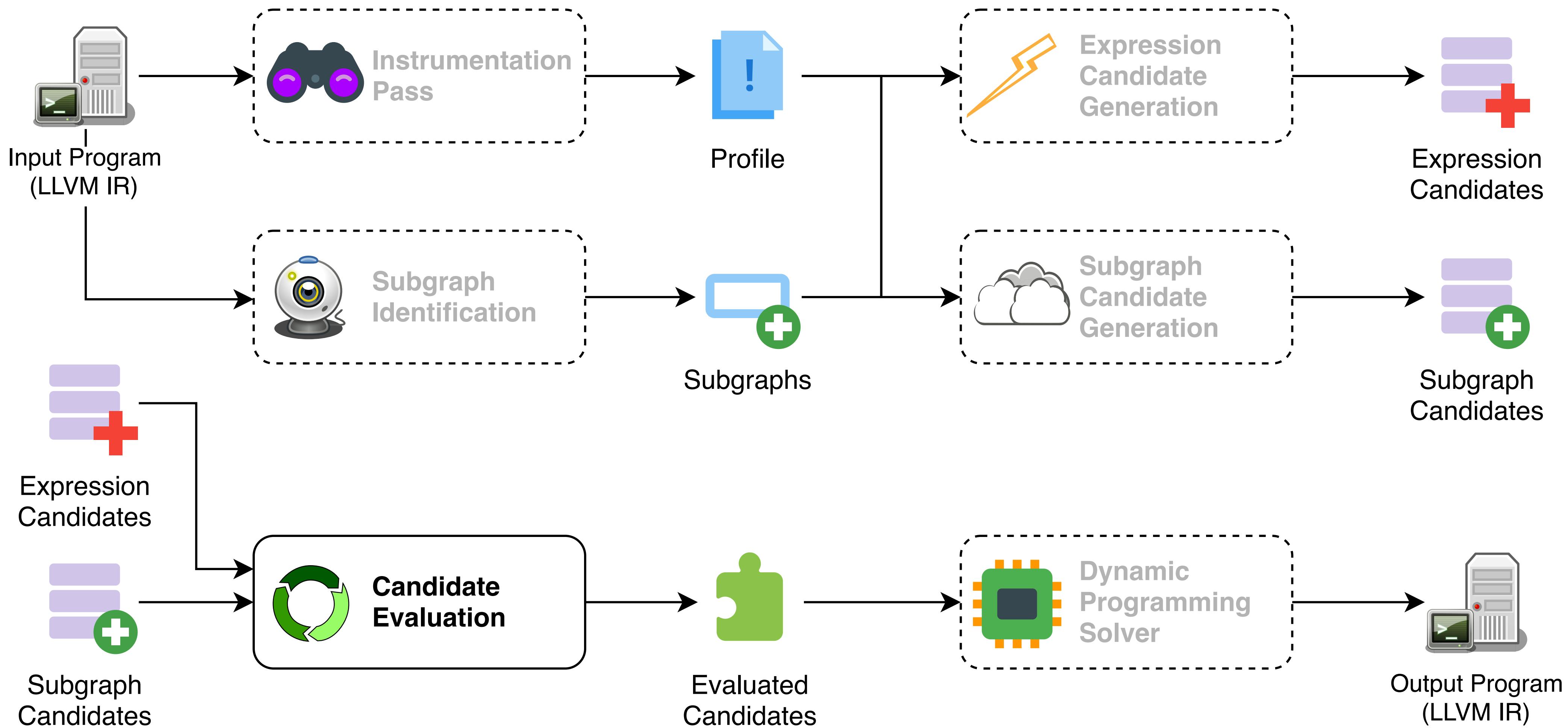


# Phase 2C: Subgraph Candidate Generation

- Gradient-Guided Precision Tuning



# Phase 2C: Candidate Evaluation



# Phase 2D: Candidate Evaluation

---

- How *local* transformations change *global* cost and accuracy?
- Cost Model
  - `llvm::TargetTransformInfo::getInstructionCost`
  - Custom Cost Model (Microbenchmarking LLVM Instructions)
    - $\text{Cost} = \sum_i \text{ExecutionCount}(i) \times \text{InstructionCost}(i)$



# Phase 2D: Candidate Evaluation

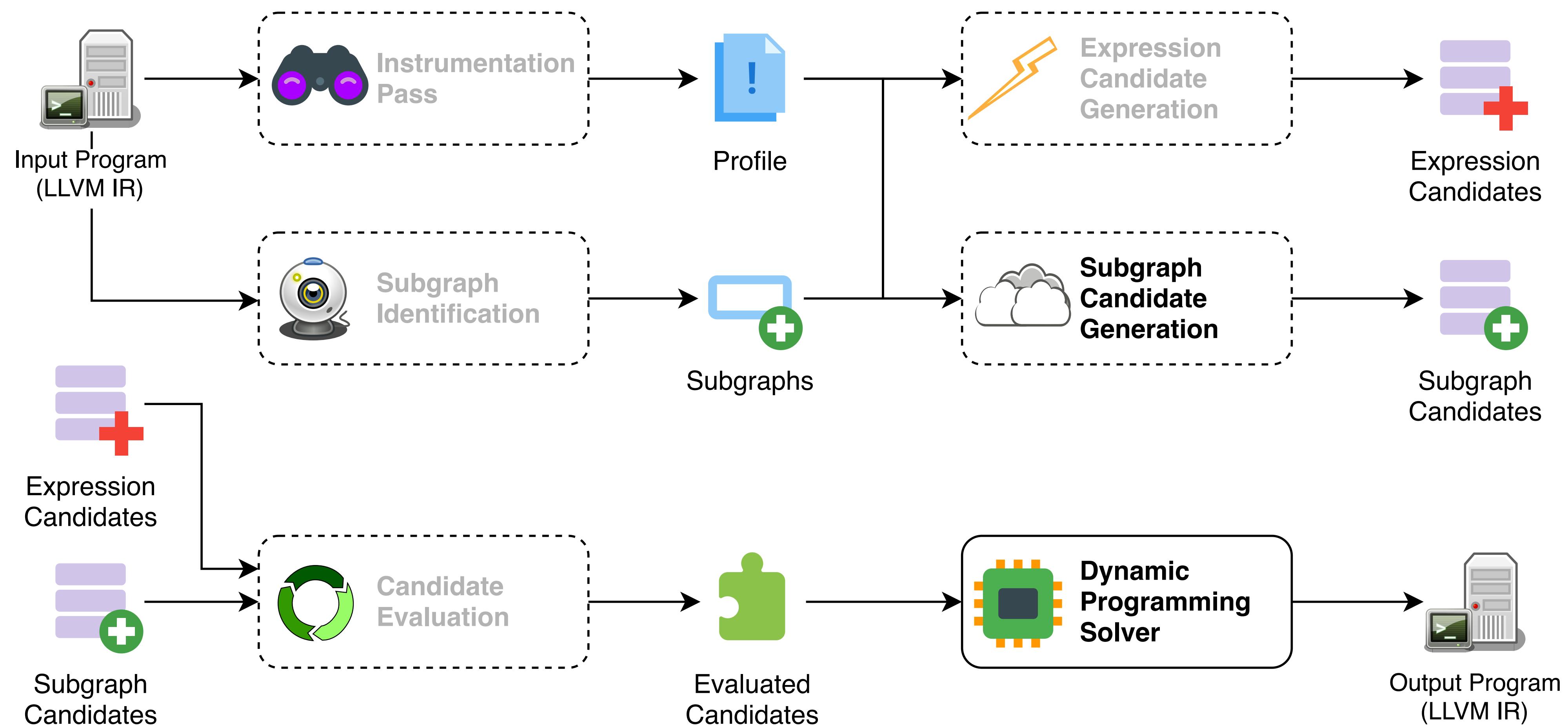
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- How *local* transformations change *global* cost and accuracy?
- Accuracy Model
  - Built-In MPFR Evaluator: **Ground Truth\*** and **Emulated Results** of *Output Instructions*
  - Global Error =  $\sum_o | \text{Gradient}(o) \times (\text{GroundTruth}(o) - \text{EmulatedResult}(o)) |$

\* Similar to techniques in Herbie (Panchekha et al. '15).



# Phase 3: Dynamic Programming Solver



# Phase 3: Dynamic Programming Solver

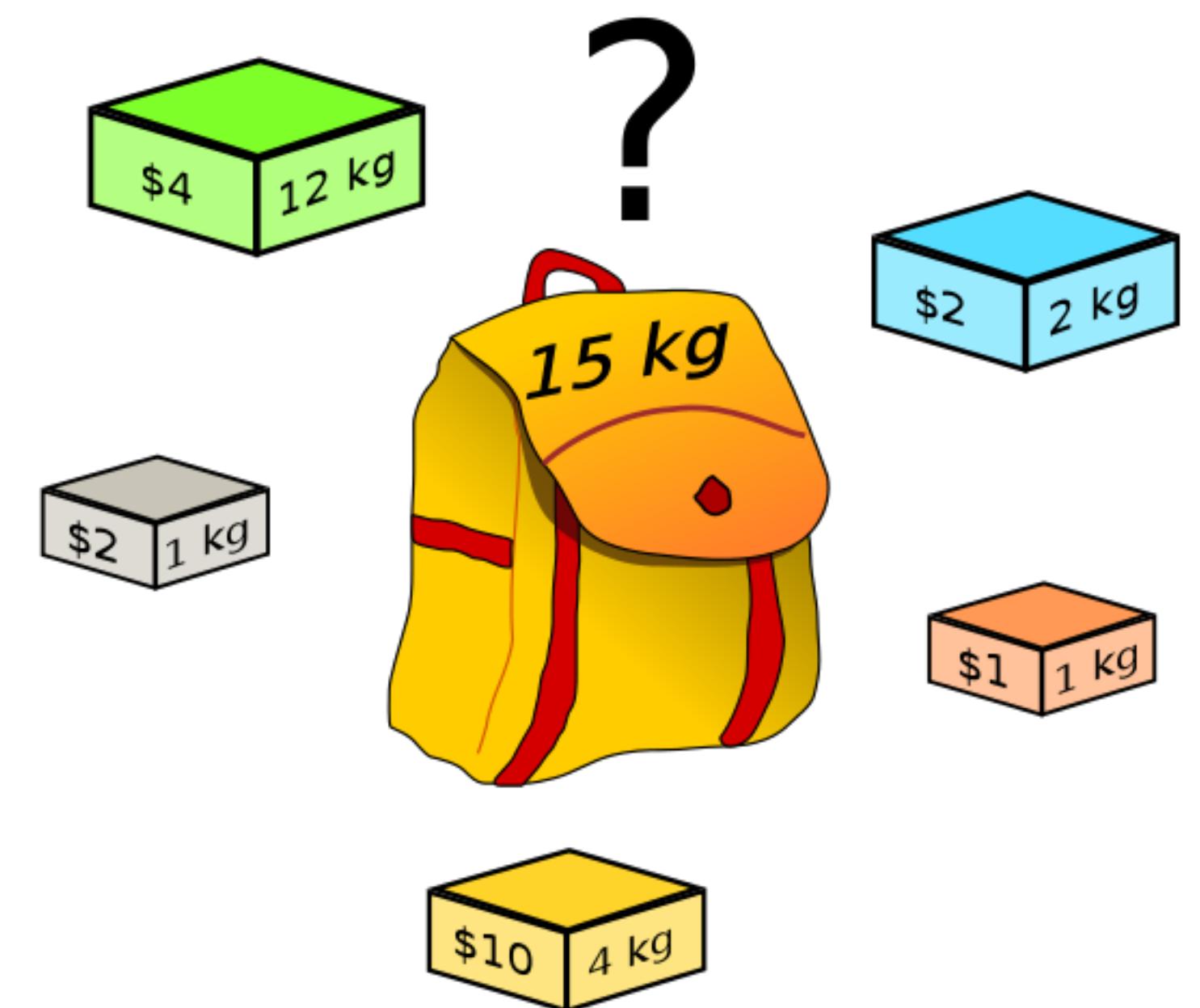
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- Poseidon's Problem
  - Known: Computation Cost **Budget**, Expression Candidates (**Cost & Global Error**), Subgraph Candidates (**Cost & Global Error**).
  - Objective:  $\operatorname{argmin}_{\phi} \sum_{\phi} \text{GlobalError}(\phi)$
  - Constraints:
    - $\sum_{\phi} \text{Cost}(\phi) \leq \text{ComputationCostBudget}$
    - Select only one candidate per  $\mathbb{F}$  expression/subgraph



# Phase 3: Dynamic Programming Solver

- 0/1 Knapsack Problem
  - Known: Knapsack **Capacity**, Items (**Weight & Profit**).
  - Objective:  $\max_{\phi} \sum_{\phi} \text{Profit}(\phi)$
  - Constraint:  $\sum_{\phi} \text{Weight}(\phi) \leq \text{Capacity}$
  - Dynamic Programming solver in pseudo-polynomial time



# Phase 3: Dynamic Programming Solver

---

- Poseidon's Problem (Rounding all costs to nearest integers)
  - Known: Computation Cost **Budget**, Expression Candidates (**Cost & Global Error**), Subgraph Candidates (**Cost\*** & **Global Error\***).
  - Objective:  $\operatorname{argmin}_{\phi} \sum_{\phi} \text{GlobalError}(\phi)$
  - Constraints:
    - $\sum_{\phi} \text{Cost}(\phi) \leq \text{ComputationCostBudget}$
    - Select only one candidate per  $\mathbb{F}$  expression/subgraph
- \* Adjusted in the solving process.



# Evaluation: FPBench Microbenchmarks

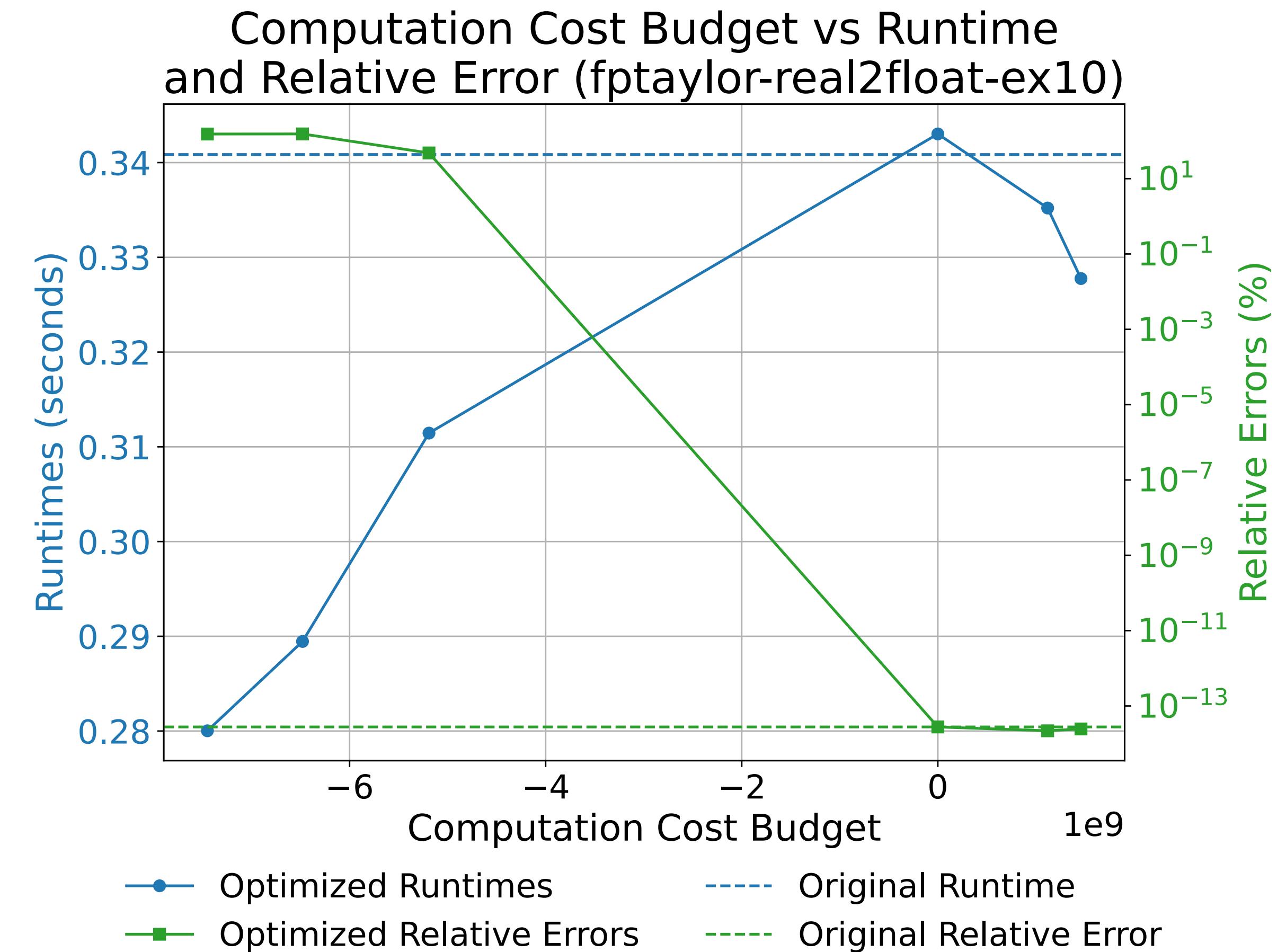
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- Small FP functions (exported to C): 46 immediately optimizable functions
- Permitting **0.0001%** Relative Error → **3.8%** Average Runtime Improvement
- Permitting **0.01%** Relative Error → **5.3%** Average Runtime Improvement
- Permitting **10%** Relative Error → **9.4%** Average Runtime Improvement
- Up to **5** bits of accuracy improvements (**0.23** bit on average)



# Evaluation: FPBench Microbenchmarks

```
double ex10(double x1, double x2, double x3, double x4, double x5, double x6) {
    return (((((x1 * x4) *
    ((((-x1 + x2) + x3) - x4) +
    x5) + x6)) + ((x2 * x5) *
    (((x1 - x2) + x3) + x4) - x5)
    + x6)) + ((x3 * x6) * (((x1
    + x2) - x3) + x4) + x5) - x6)))
    - ((x2 * x3) * x4)) - ((x1 *
    x3) * x5)) - ((x1 * x2) * x6))
    - ((x4 * x5) * x6);
}
```

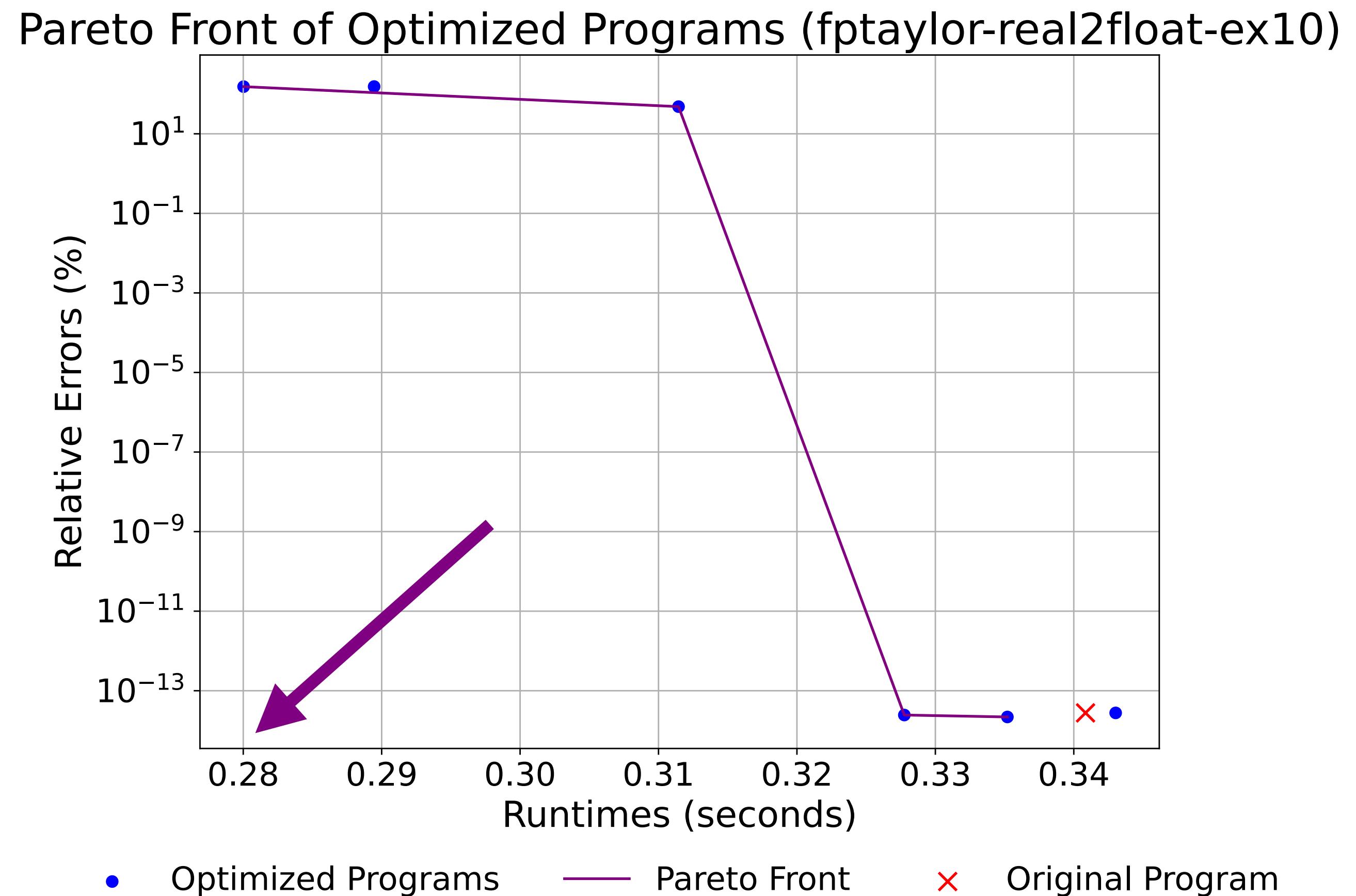


Lower is better



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    + x2) - x3) + x4) + x5) - x6)))
    - ((x2 * x3) * x4)) - ((x1 *
    x3) * x5)) - ((x1 * x2) * x6))
    - ((x4 * x5) * x6);
```





# Poseidon (Open Source on GitHub: [EnzymeAD/Enzyme](#))

- An end-to-end system that automatically performs profile-guided two-phase optimizations of floating-point programs in compilers
- **Instrumentation Pass**
- **Expression/Subgraph Transformation Candidate Generation & Evaluation**
- **Dynamic Programming Solver**
- Permitting **0.01%** Relative Error → **5.3%** Average Runtime Improvement

