

### Synthesize Solving Strategy for Symbolic Execution

Zhenbang Chen

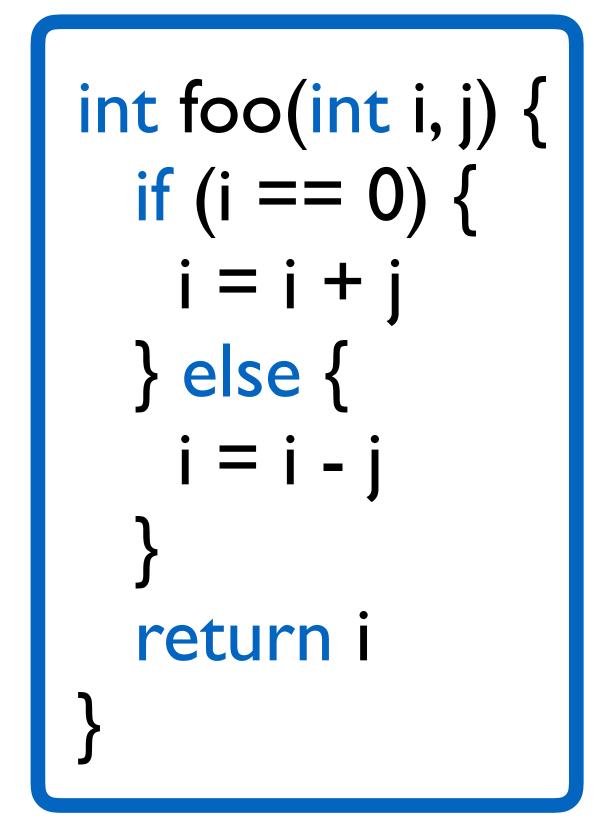
(<u>zbchen@nudt.edu.cn</u>)

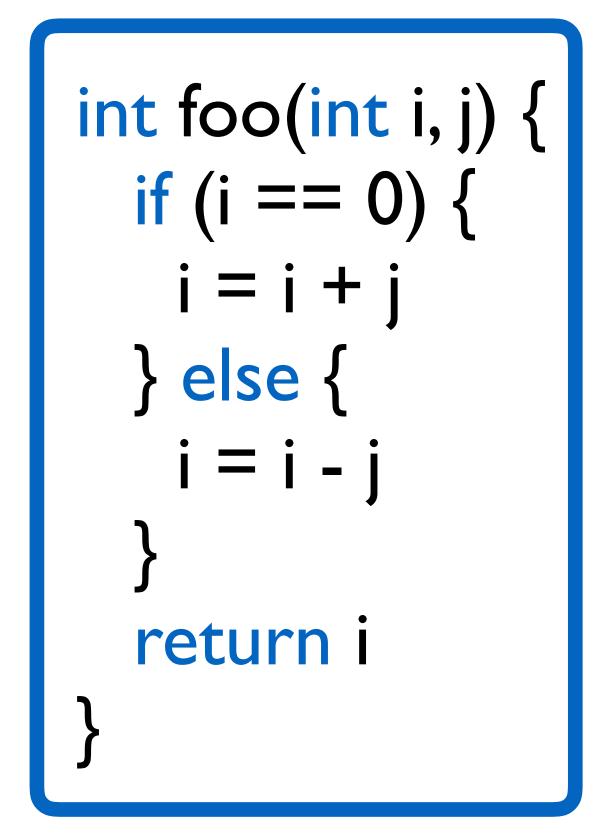
Joint work with Zehua Chen, Ziqi Shuai, Guofeng Zhang, Weiyu Pan, Yufeng Zhang, and Ji Wang



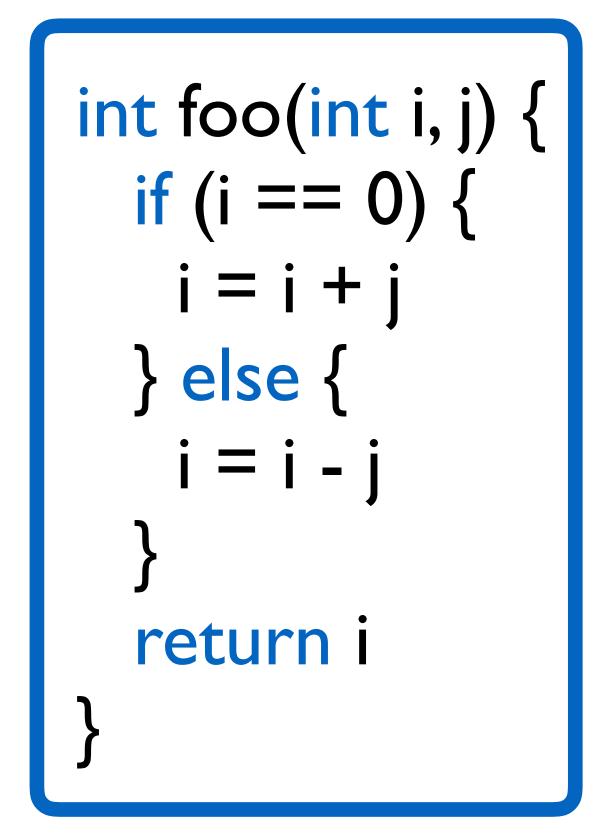
ACM SIGSOFT International Symposium on Software Testing and Analysis







$$i, j \leftarrow x_i, x_j$$
$$x_i = 0$$
$$i, j \leftarrow x_i + x_j, x_j$$
$$ret \leftarrow x_i + x_j$$



$$i, j \leftarrow \mathbf{x}_i, \mathbf{x}_j$$

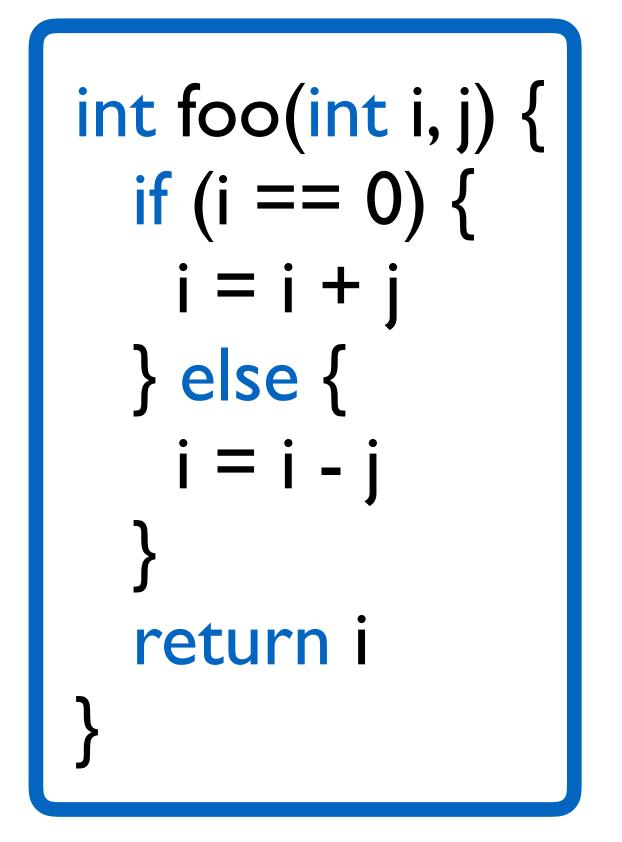
$$x_i = 0 \qquad x_i \neq 0$$

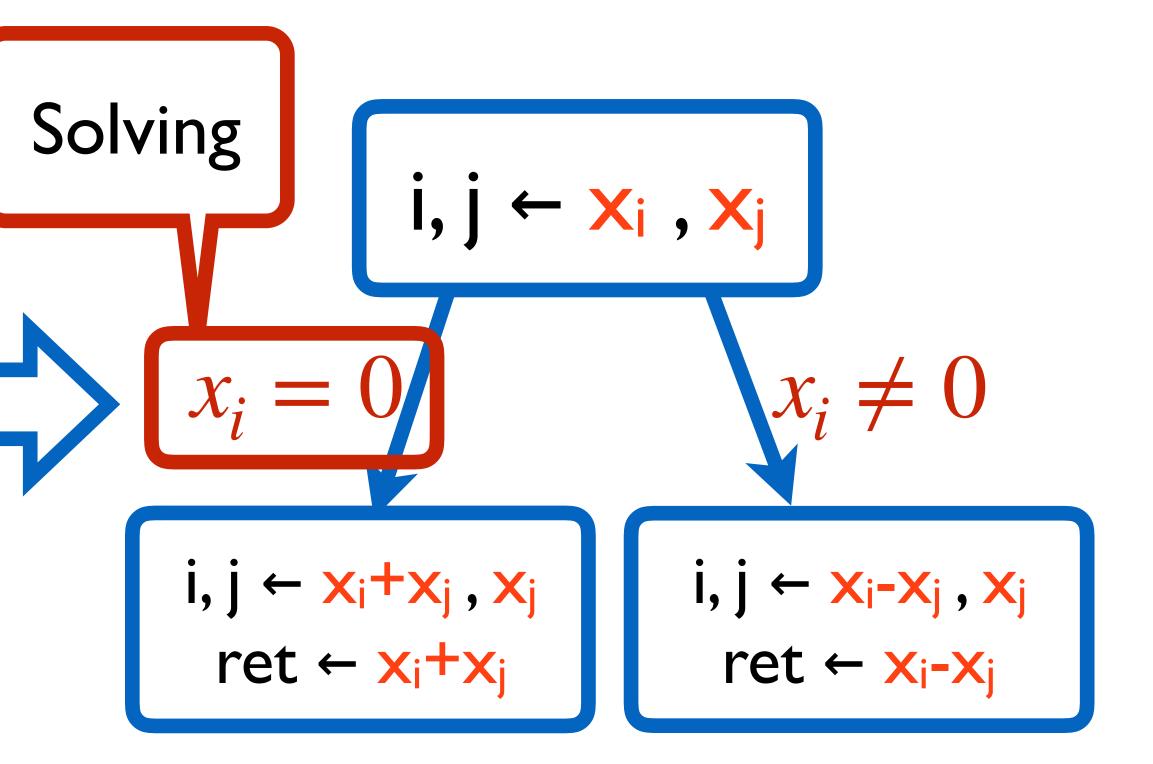
$$i, j \leftarrow \mathbf{x}_i + \mathbf{x}_j, \mathbf{x}_j$$

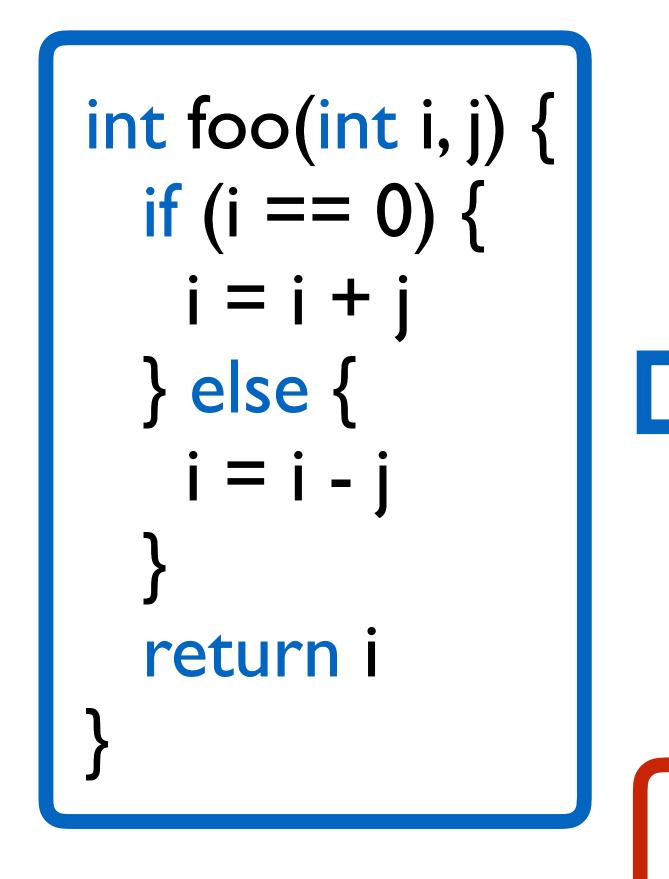
$$ret \leftarrow \mathbf{x}_i + \mathbf{x}_j$$

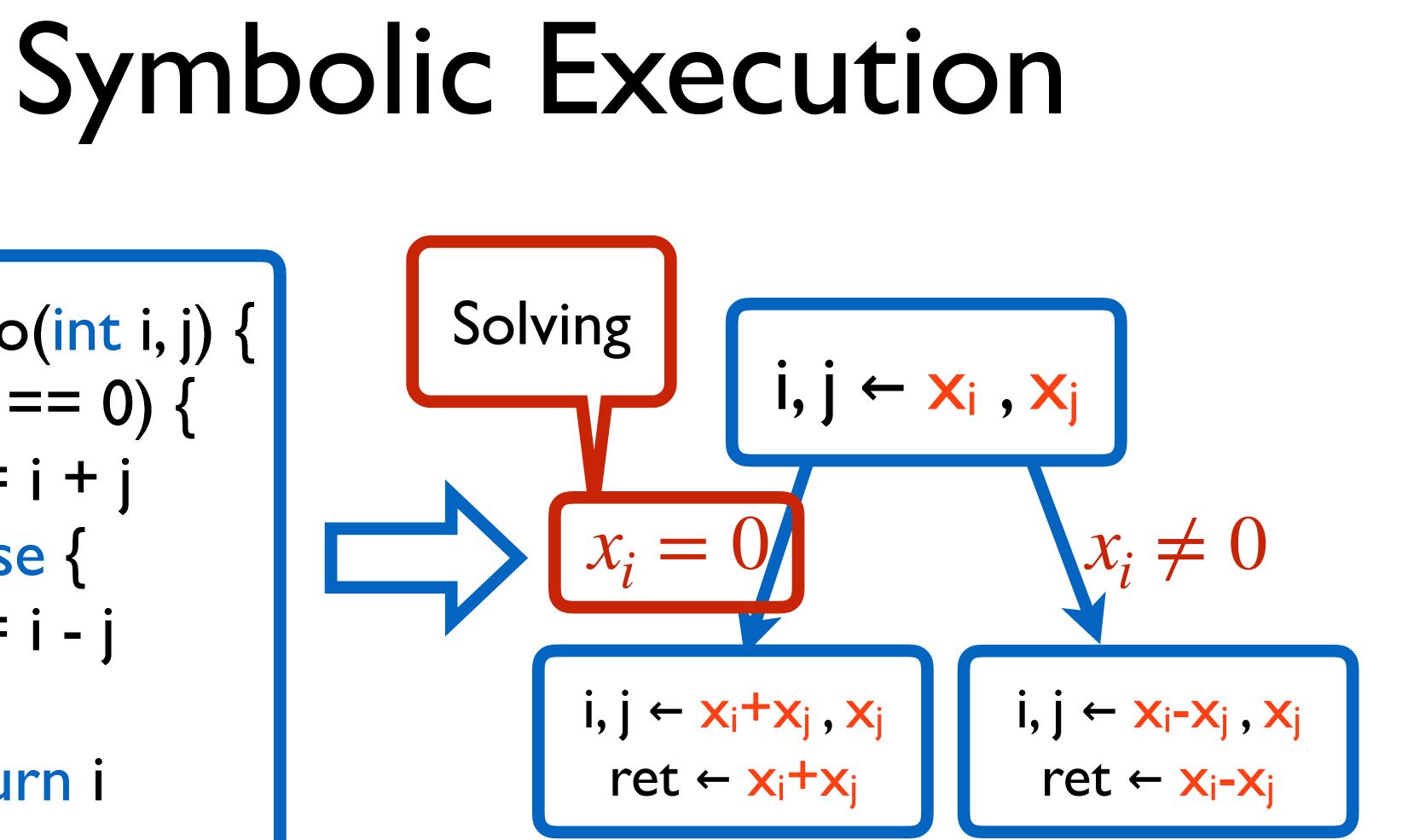
$$i, j \leftarrow \mathbf{x}_i - \mathbf{x}_j, \mathbf{x}_j$$

$$ret \leftarrow \mathbf{x}_i - \mathbf{x}_j$$









Constraint solving is the enabling technique

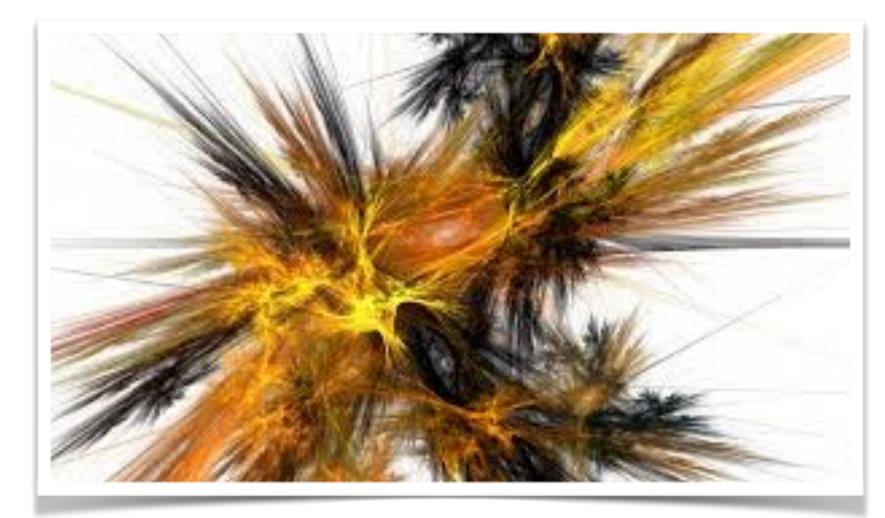


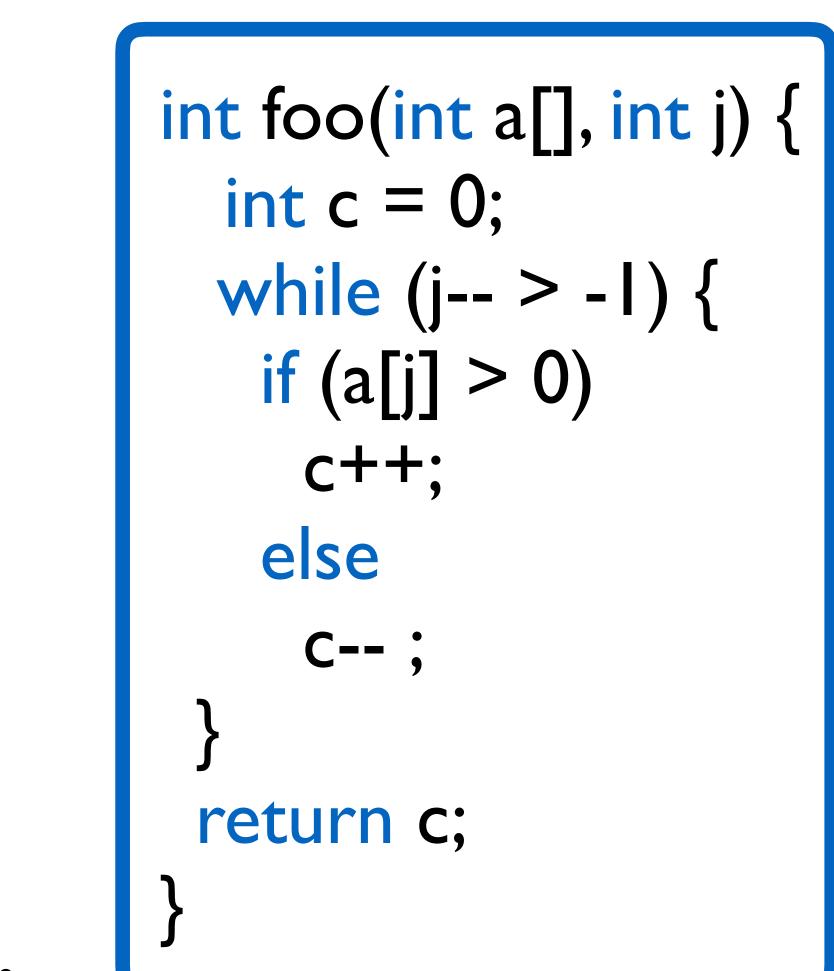
#### Path explosion



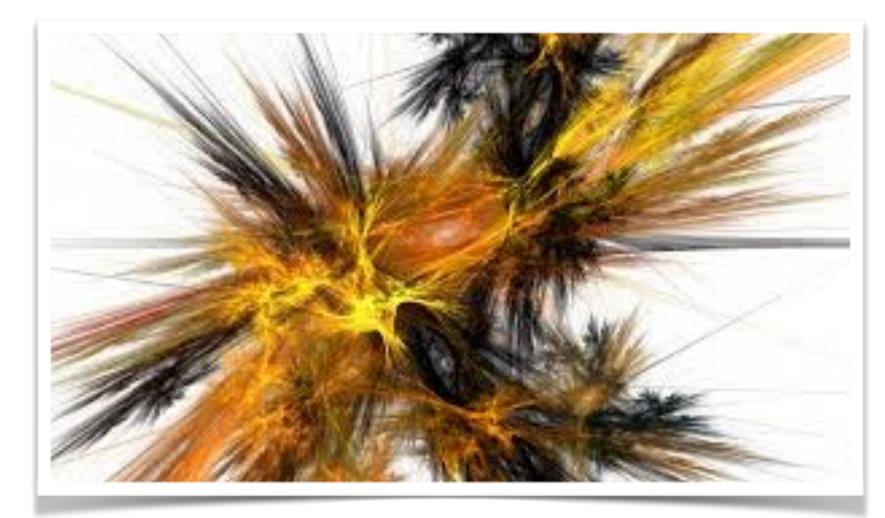


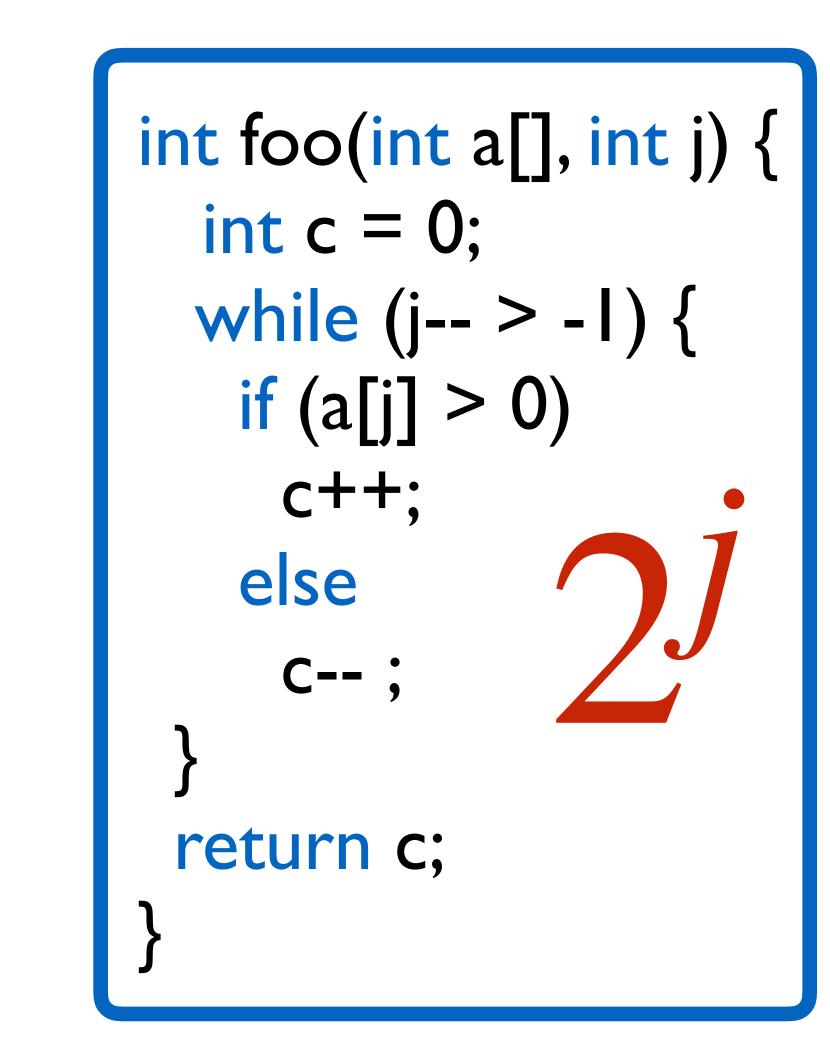
#### Path explosion



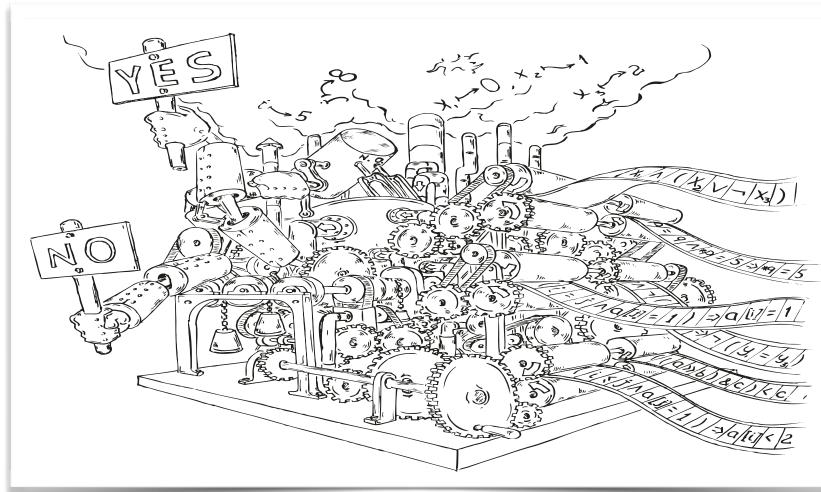


#### Path explosion





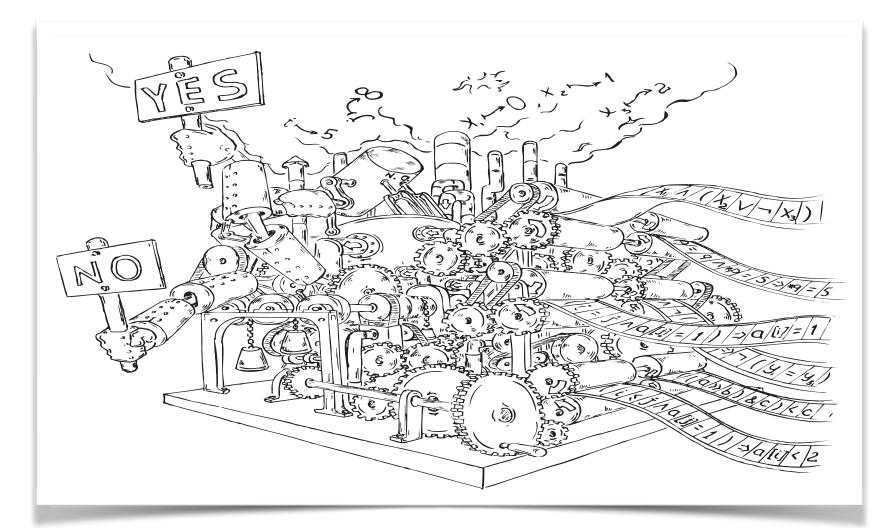
#### Constraint Solving



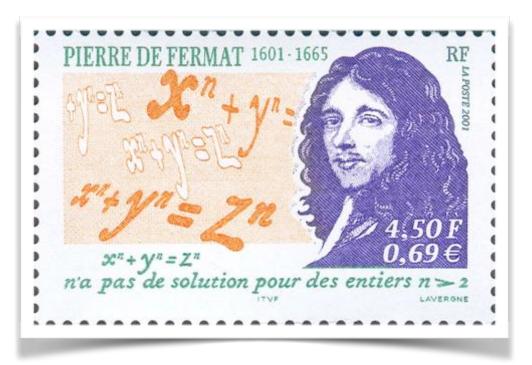
Decision Procedures An Algorithmic Point of View, Second Edition, 2016



#### Constraint Solving



Decision Procedures An Algorithmic Point of View, Second Edition, 2016



Undecidable in general
 High complexity in computation



Decision Procedures An Algorithmic Point of View, Second Edition, 2016

## Our Work's Target



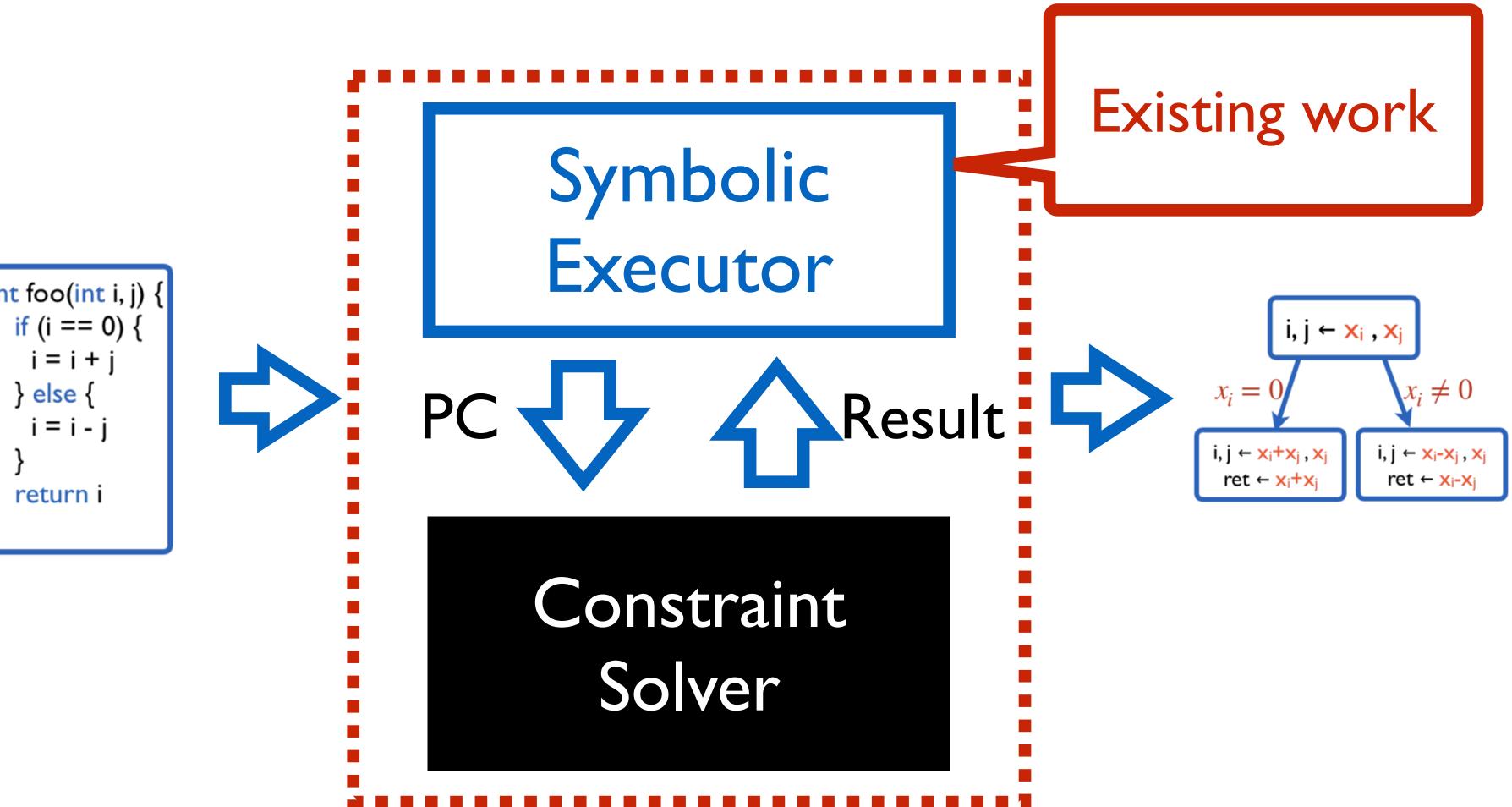
Decision Procedures An Algorithmic Point of View, Second Edition, 2016

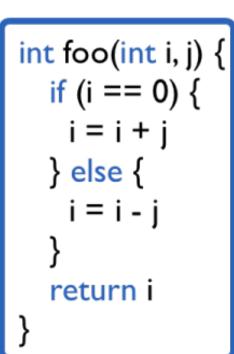
- Query cache (partial) and simplification
  - KLEE[OSDI'08], KLEE-Array[ISSTA'17]

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  - KLEE[OSDI'08], KLEE-Array[ISSTA'17]
- Query reduction
  - SSE[ISSRE'12], Cloud9[PLDI'12]

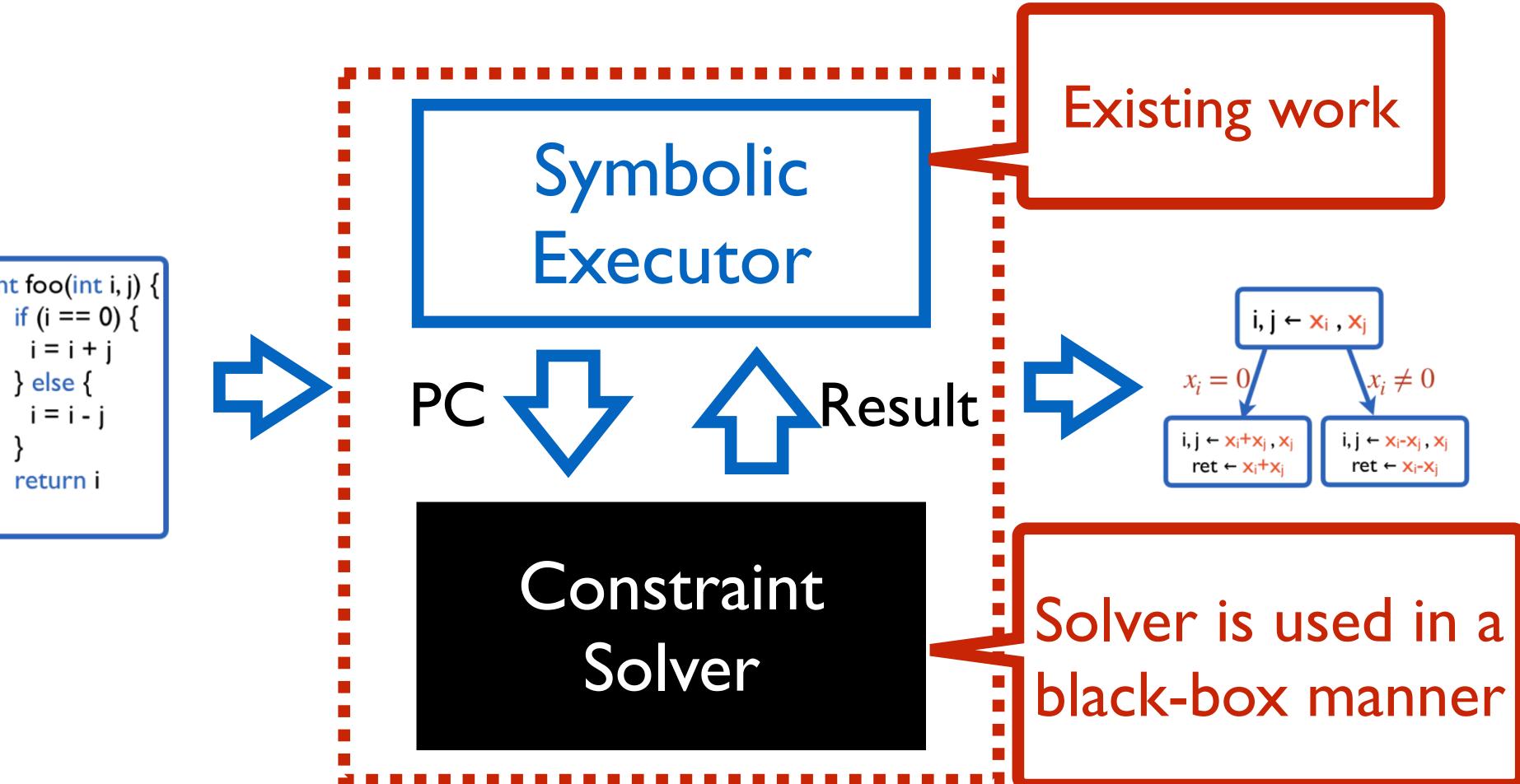
- Query cache (partial) and simplification
  - KLEE[OSDI'08], KLEE-Array[ISSTA'17]
- Query reduction
  - SSE[ISSRE'12], Cloud9[PLDI'12]
- Query reuse
  - Green[FSE'I2], GreenTrie[ISSTA'I5]

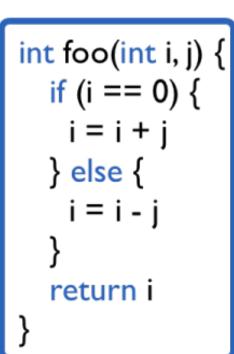
### Our Observation

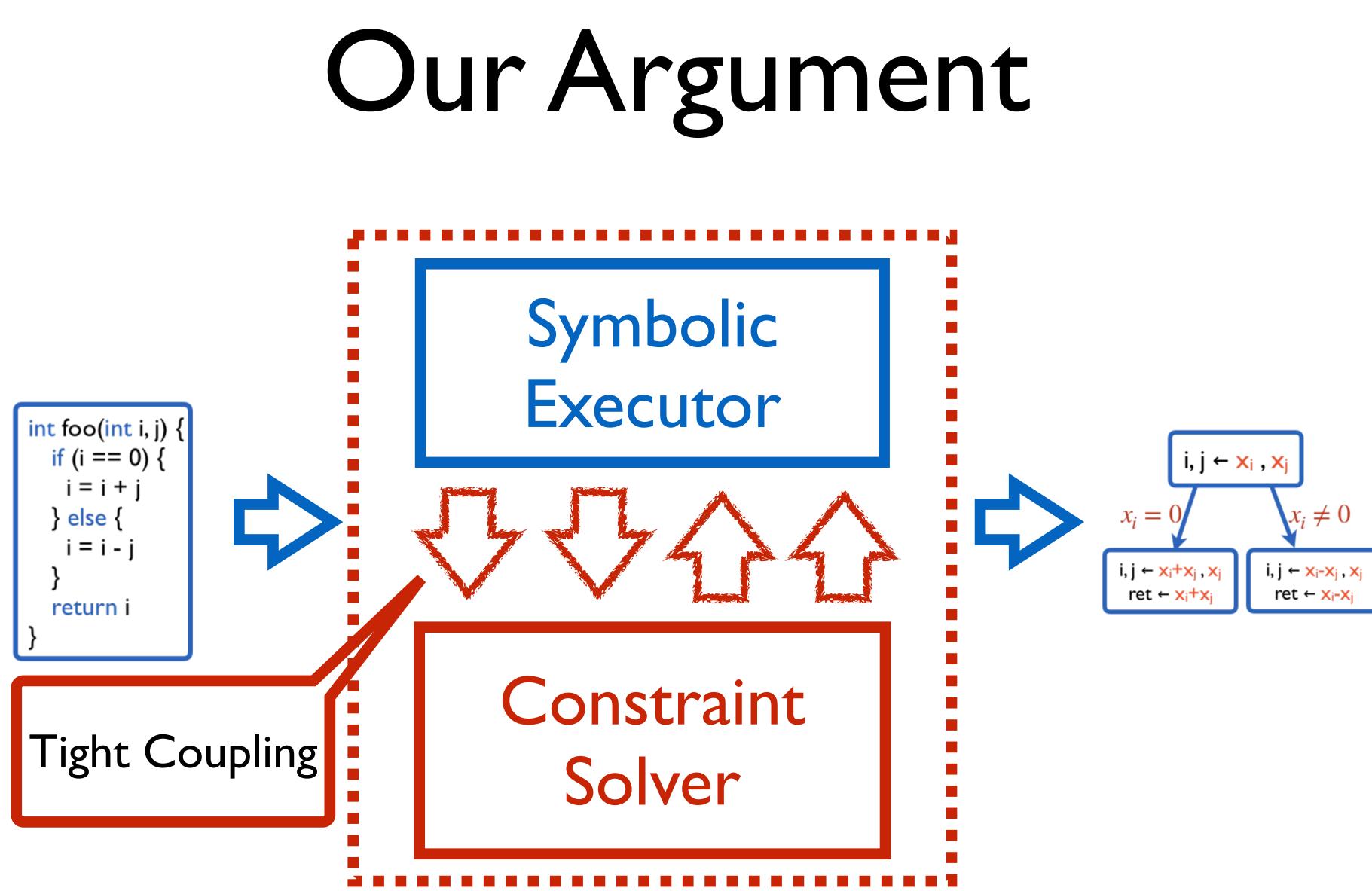


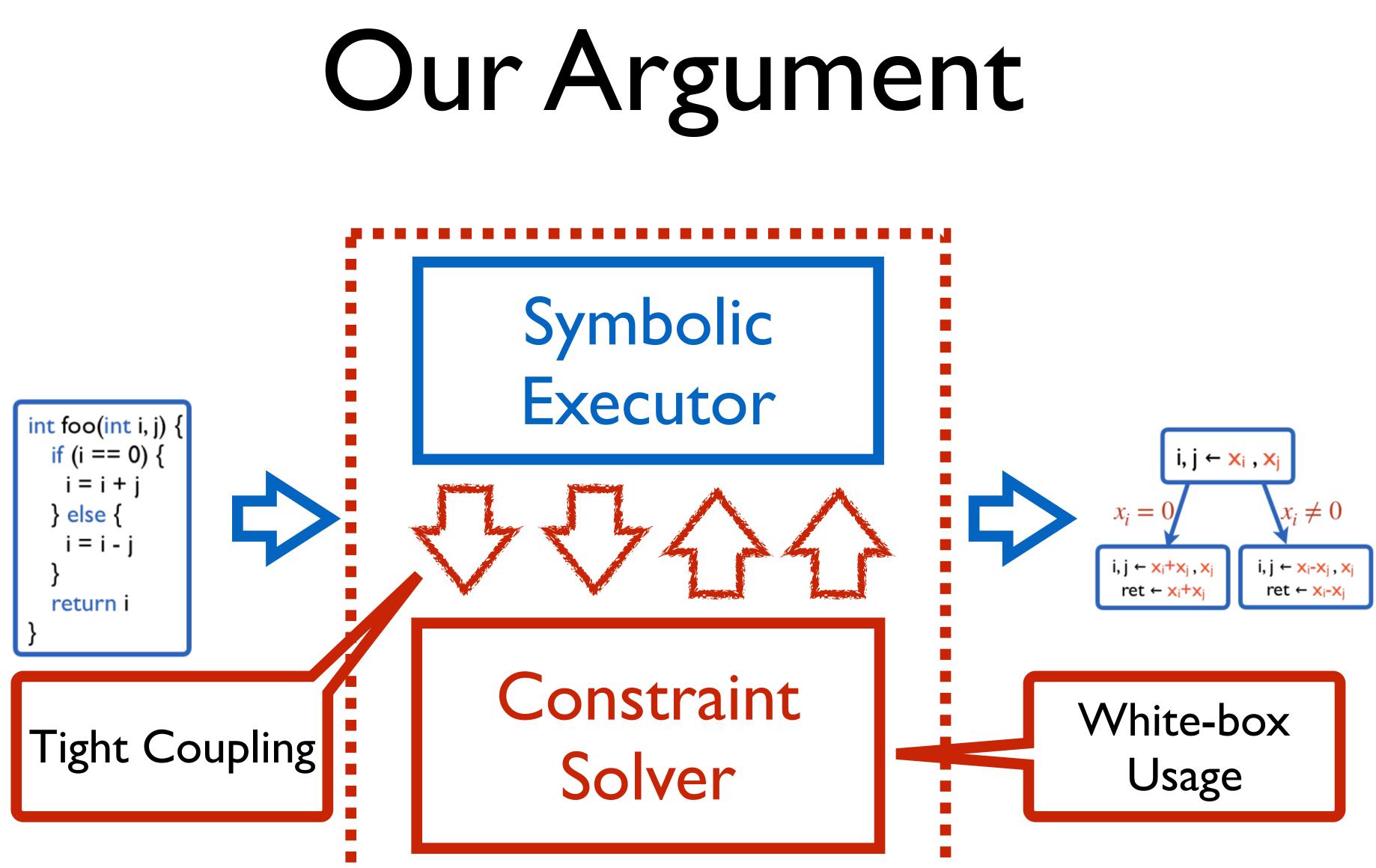


### Our Observation



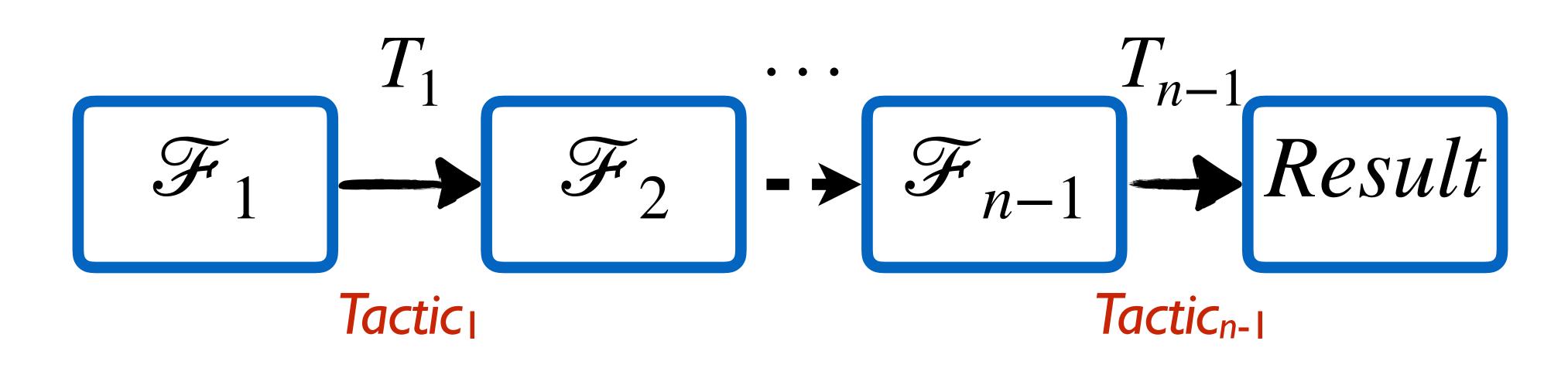




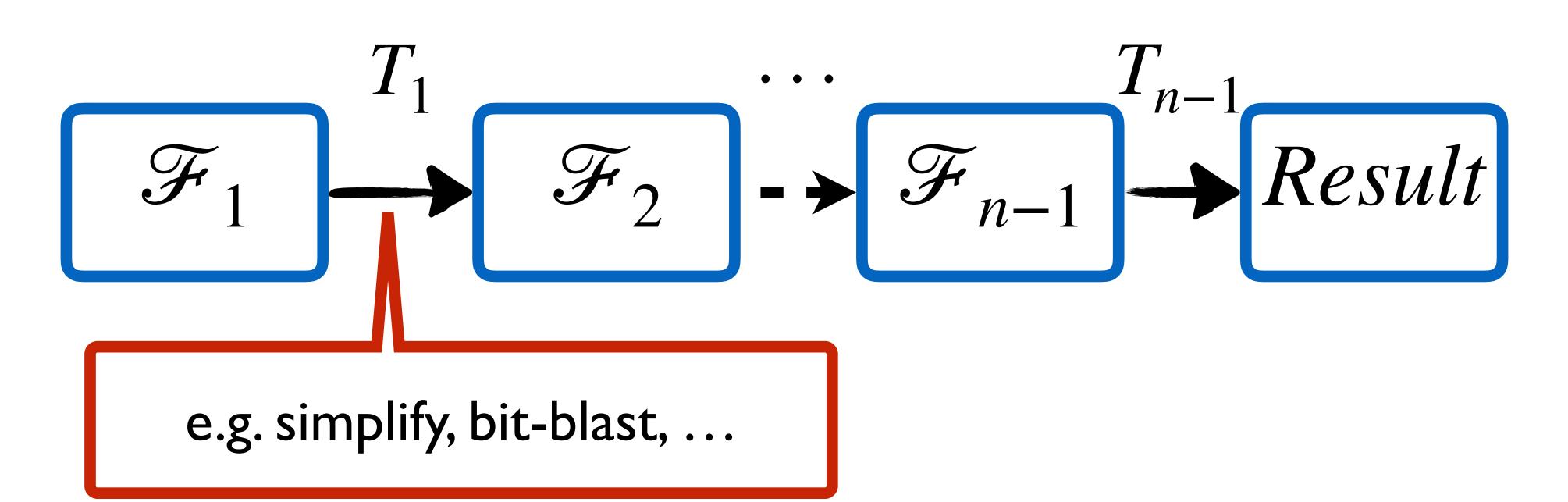


#### • Customize the solving procedure

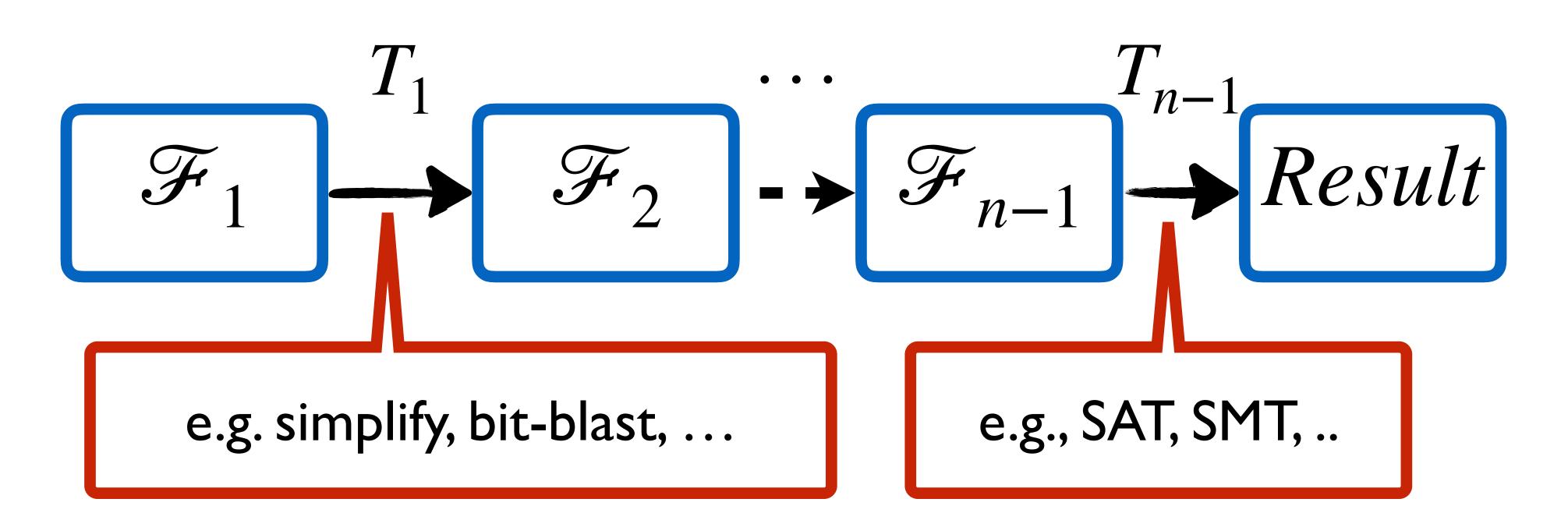
#### Customize the solving procedure

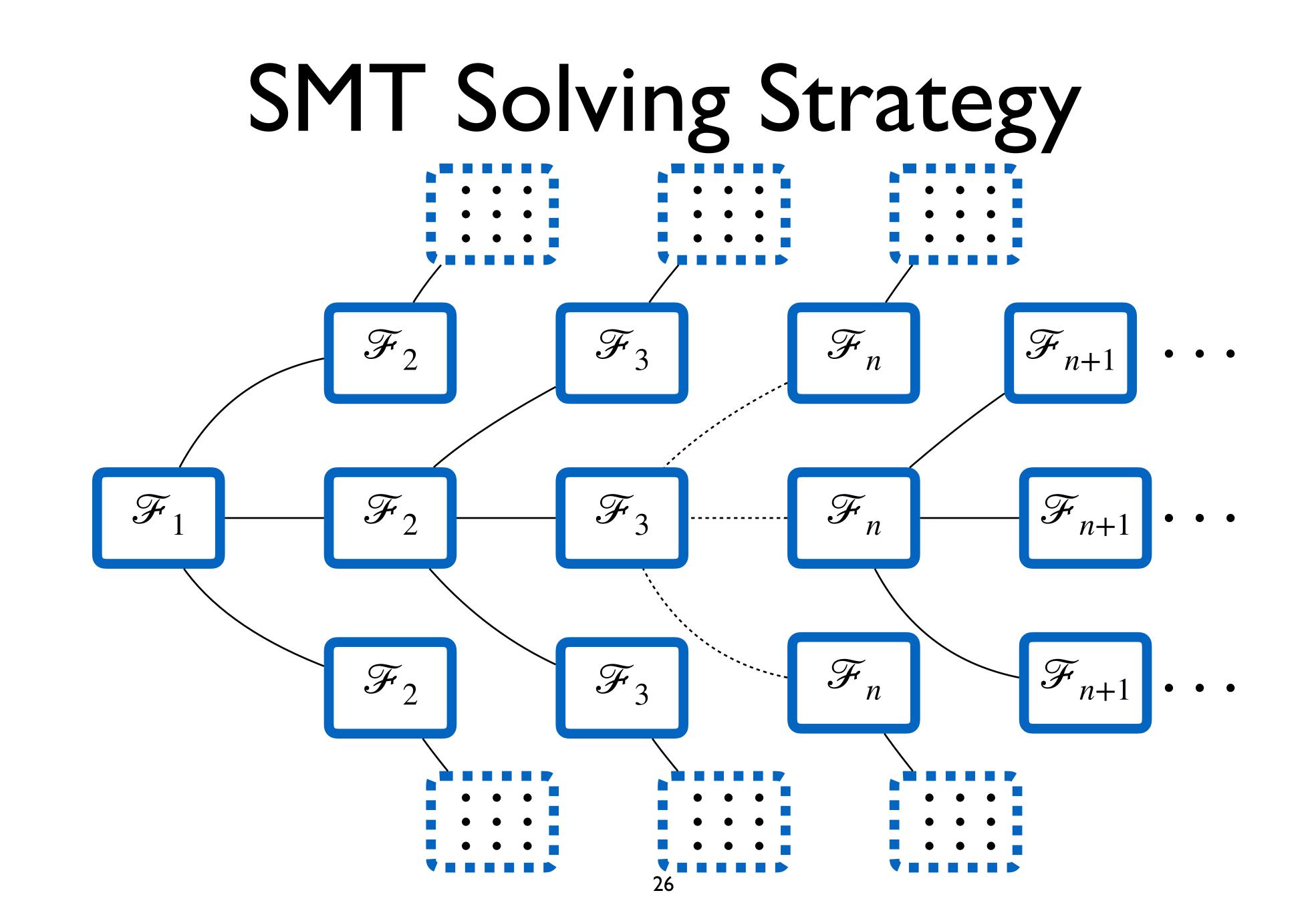


#### Customize the solving procedure



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#### • Solving strategy has a great influence to solving performance

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### $x^3 = 8.0$ QF\_BVFP formula, x is a double variable

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 $x^3 = 8.0$ 

56 seconds by using Z3's default strategy

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#### QF\_BVFP formula, x is a double variable

 $x^3 = 8.0$ 

56 seconds by using Z3's default strategy

22 seconds by using the following customized one *(simplify, SMT)* 

## Our Key Insight

### A program's symbolic execution is a specific constraint solving problem

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- problem

• A program's symbolic execution is a specific constraint solving

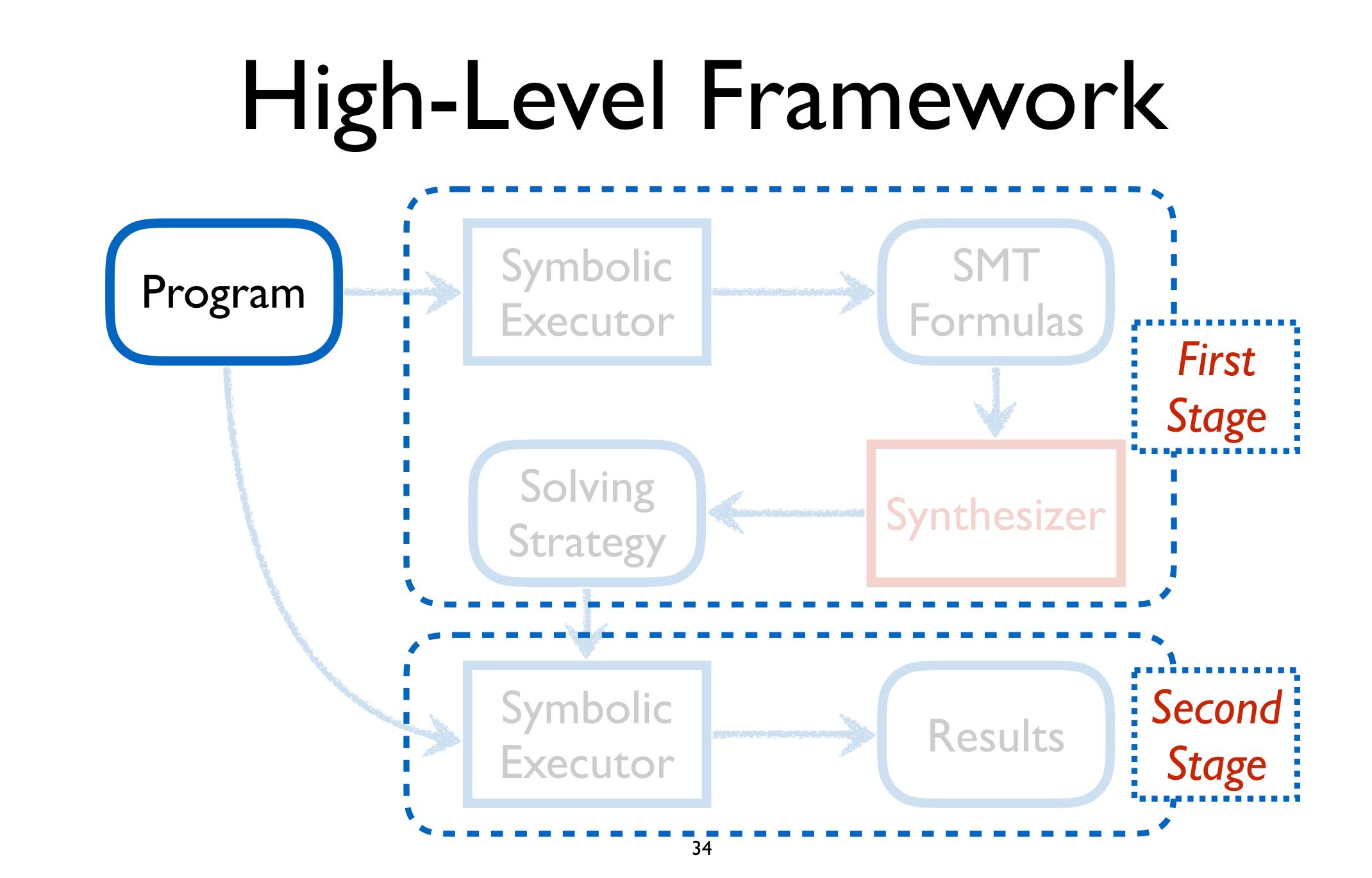
• We can use solving strategy to customize the solver for the program to solve the program's path constraints efficiently

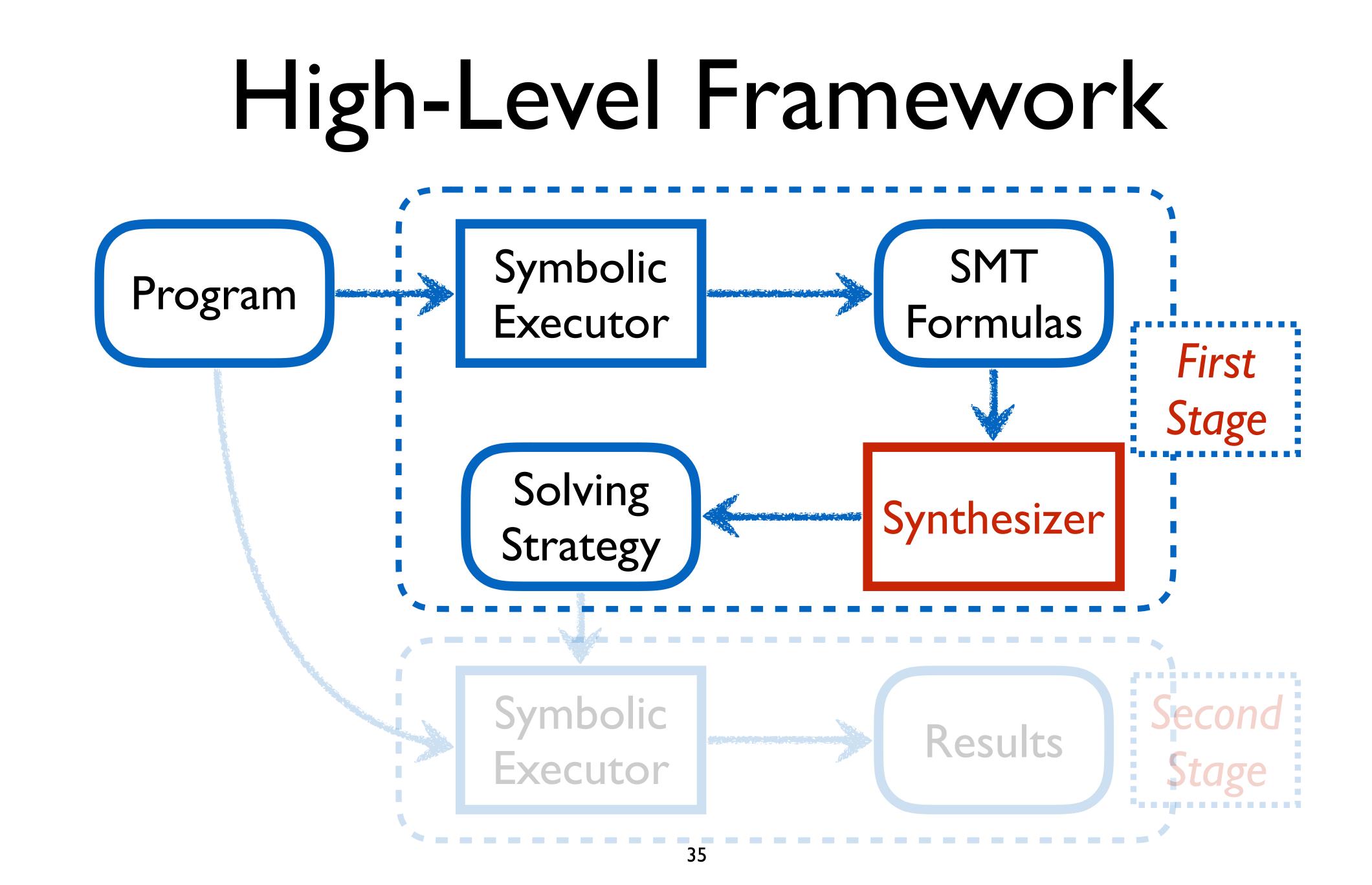
### Our Key Idea

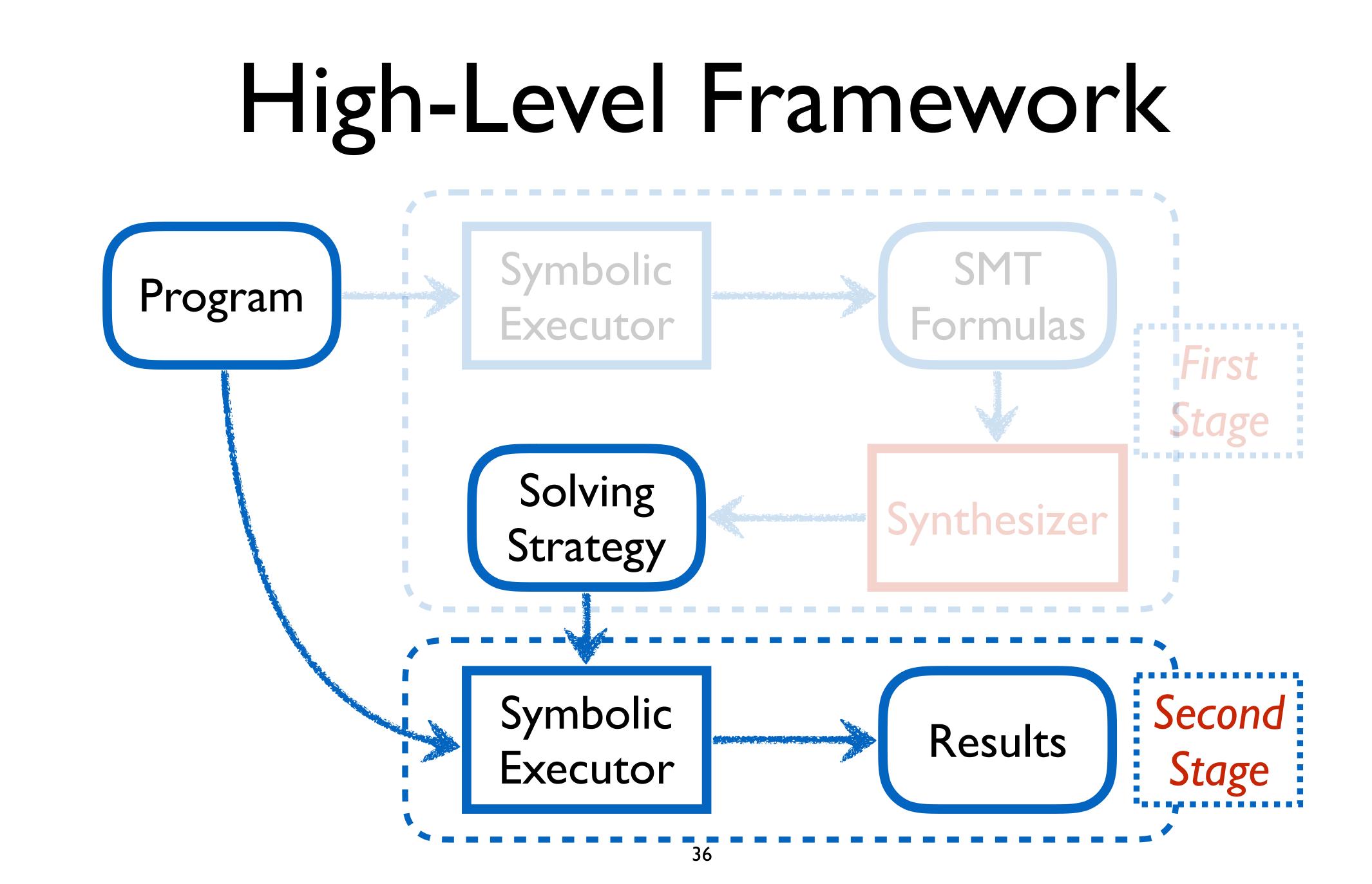
- symbolic execution

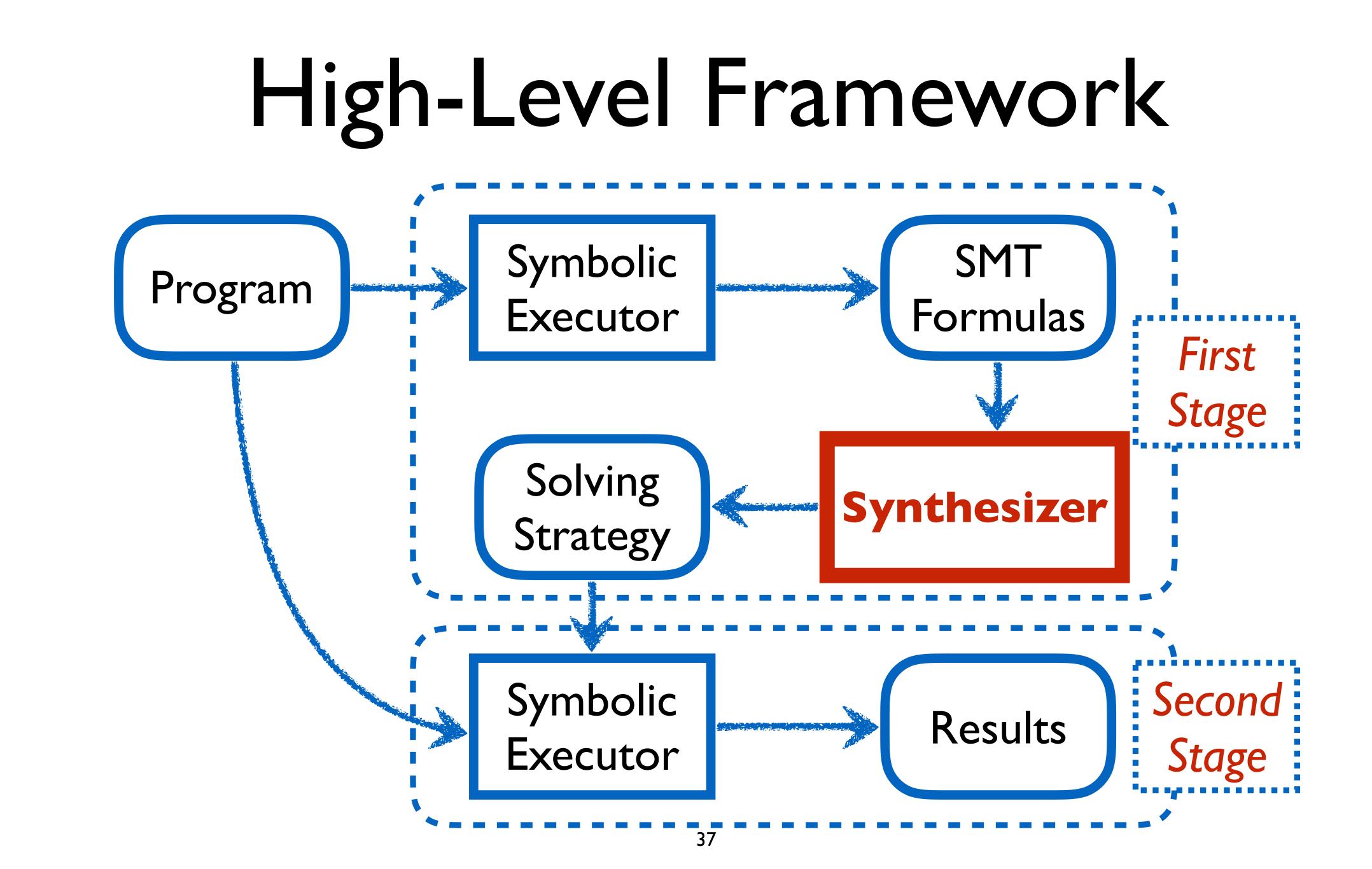
#### • Online synthesize a solving strategy for the program under

#### The synthesized solving strategy can improve the efficiency of solving the program's path conditions in symbolic execution

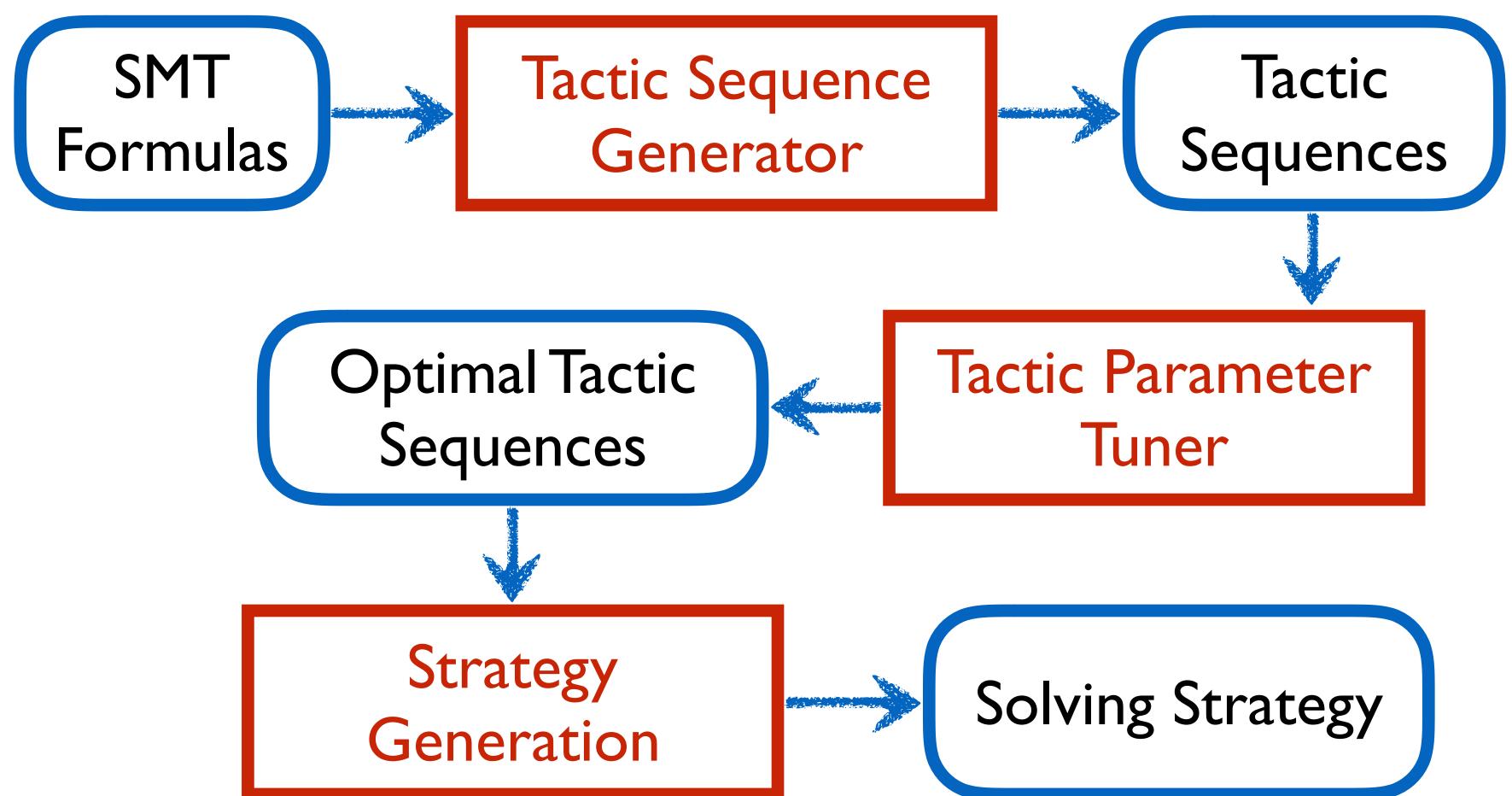






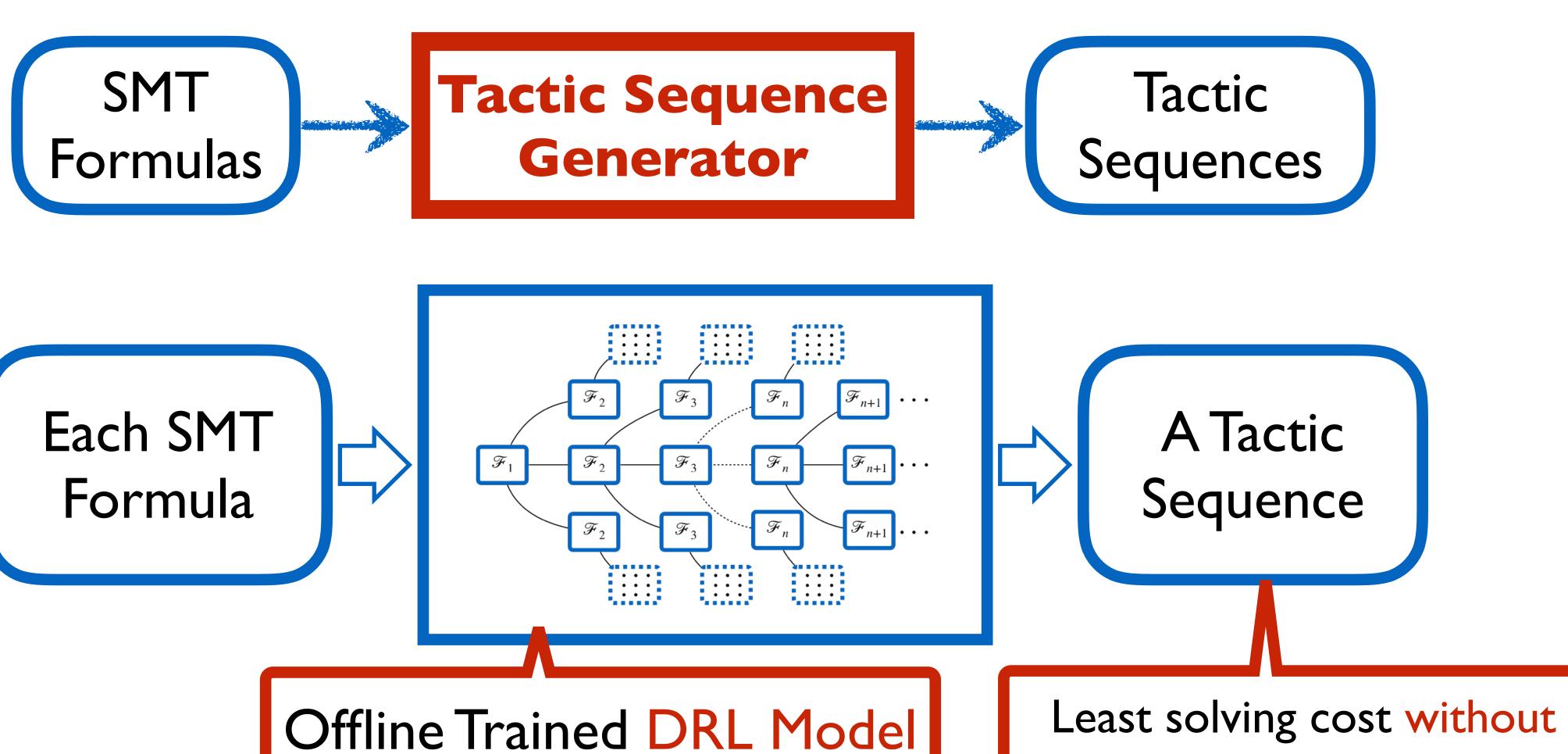


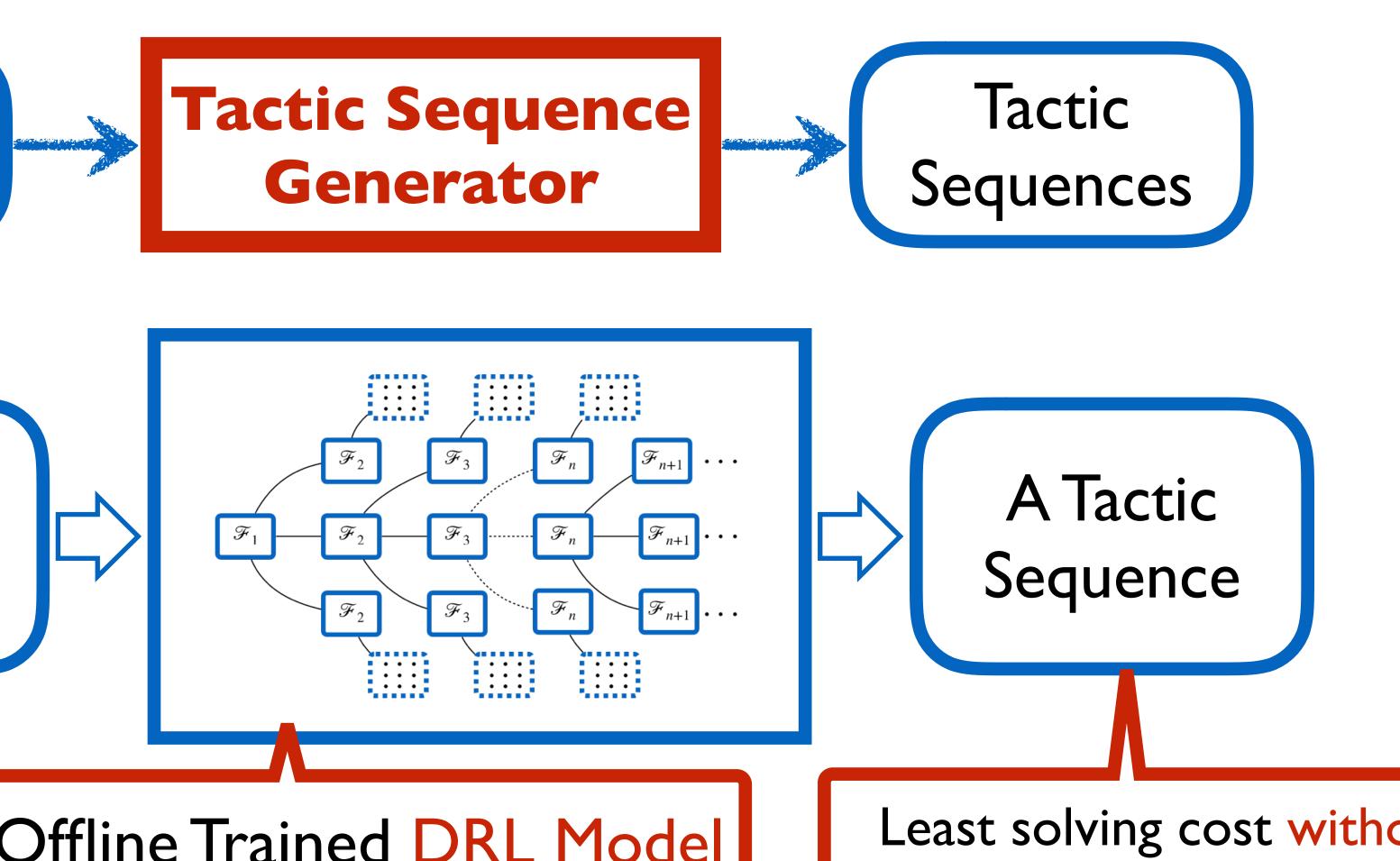
# Synthesizer's Details





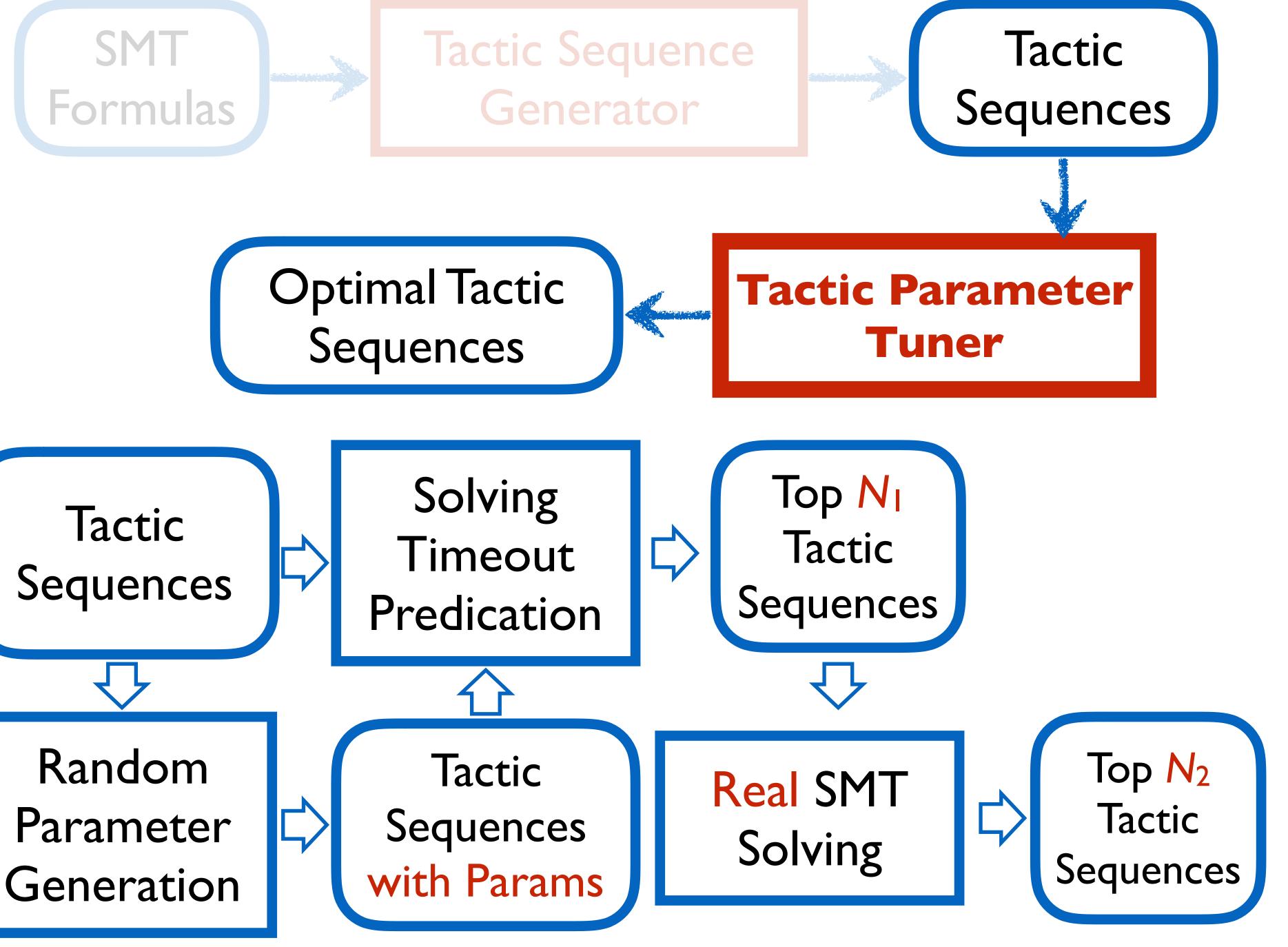
# Synthesizer's Details

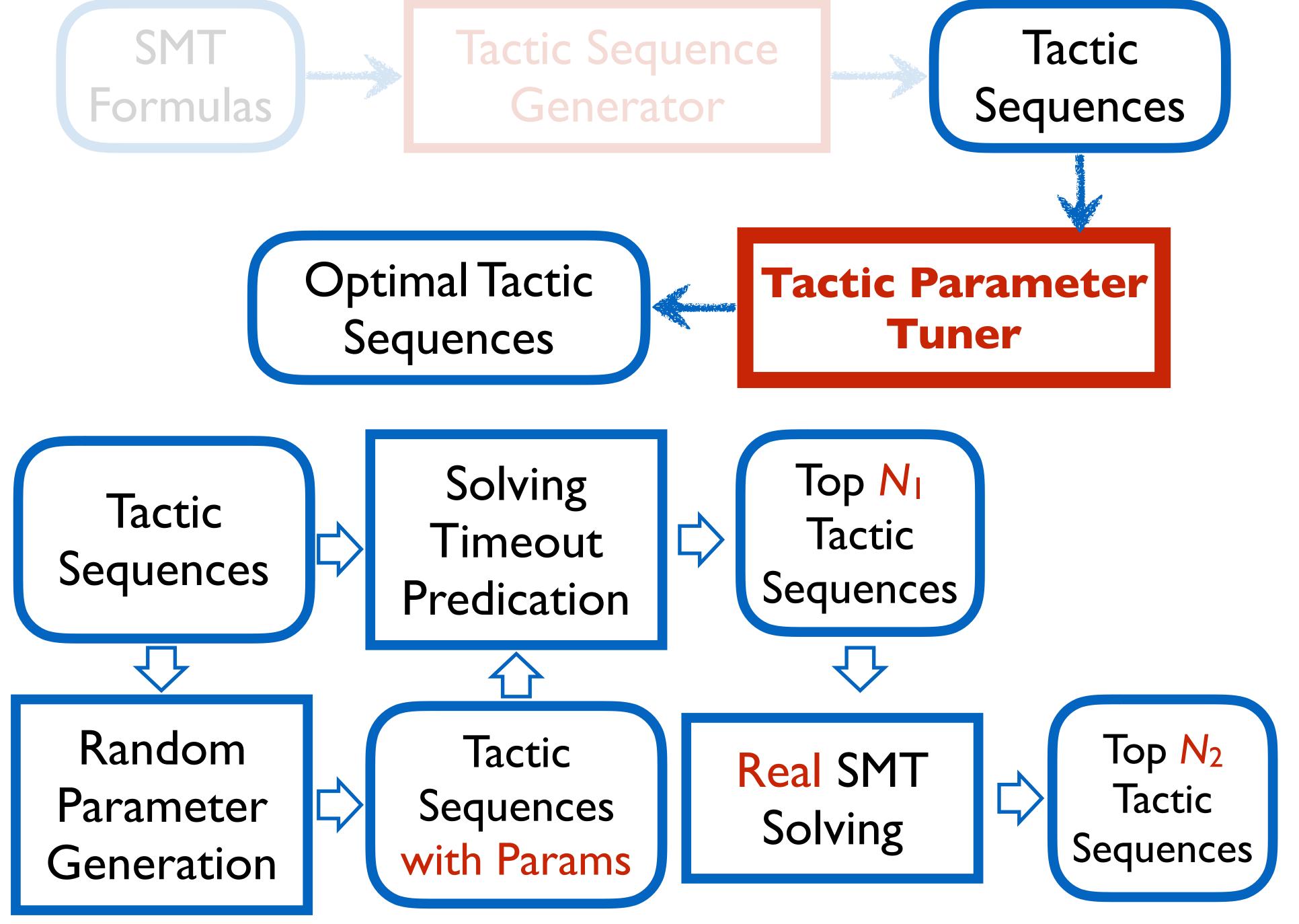


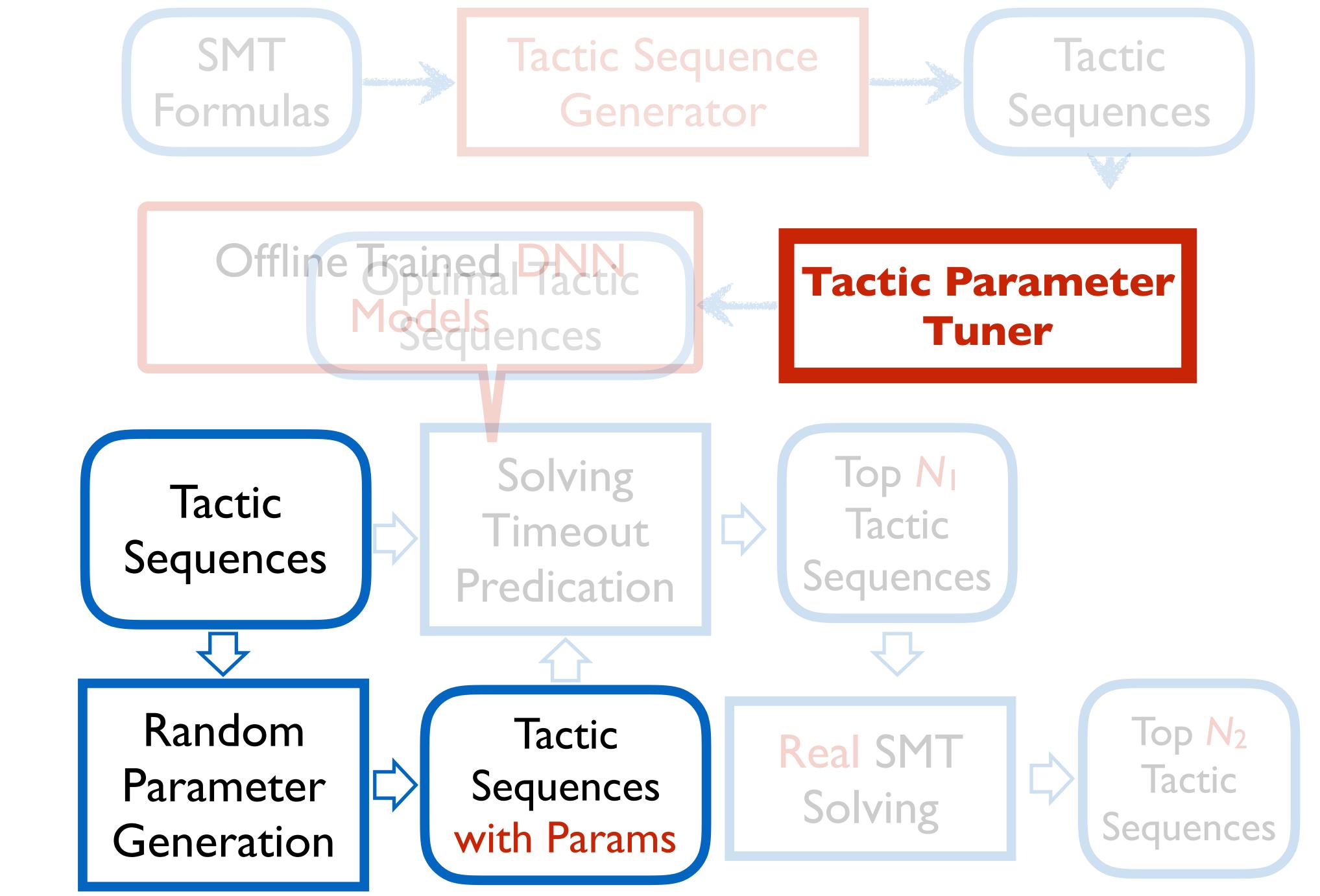


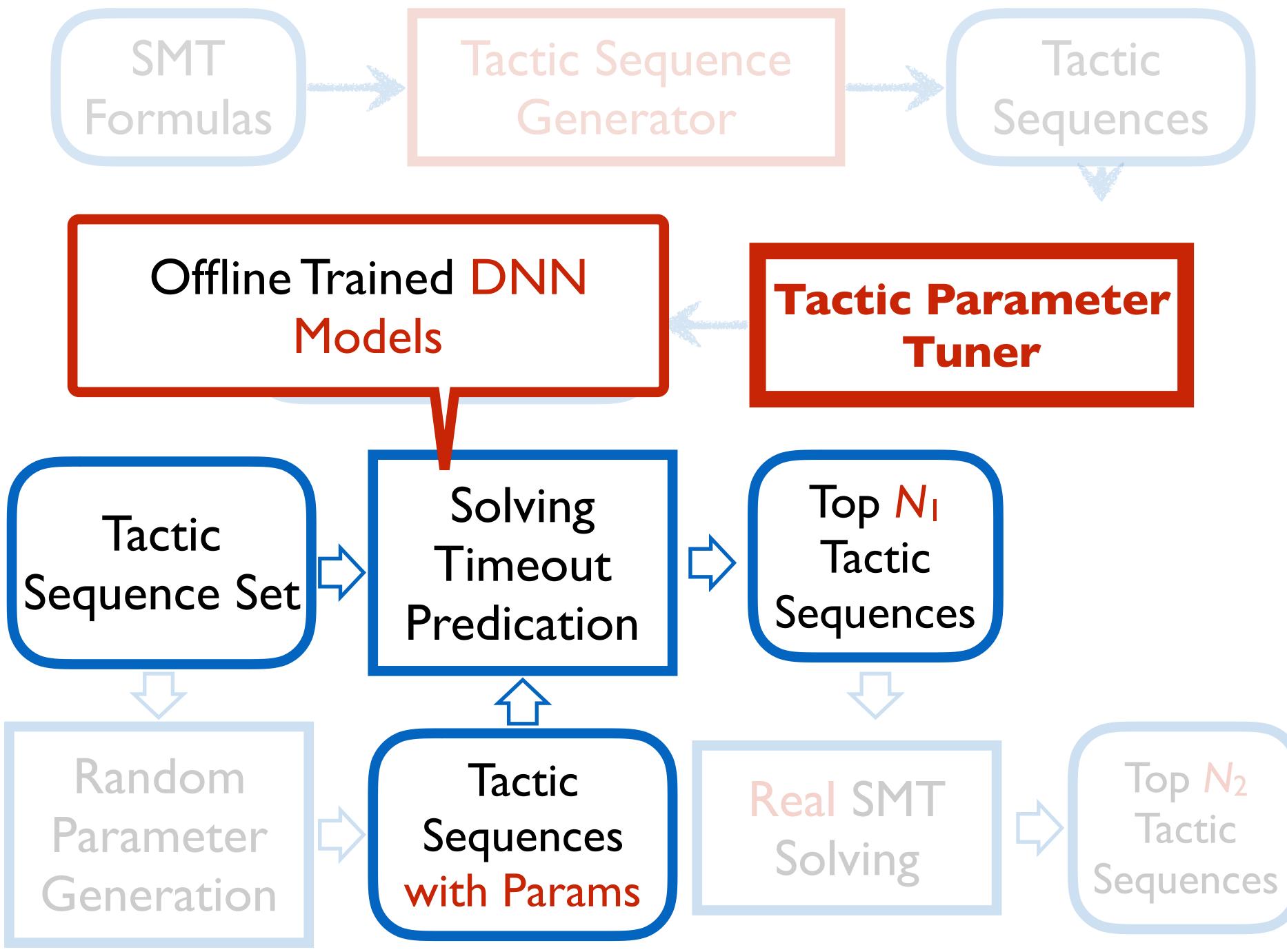


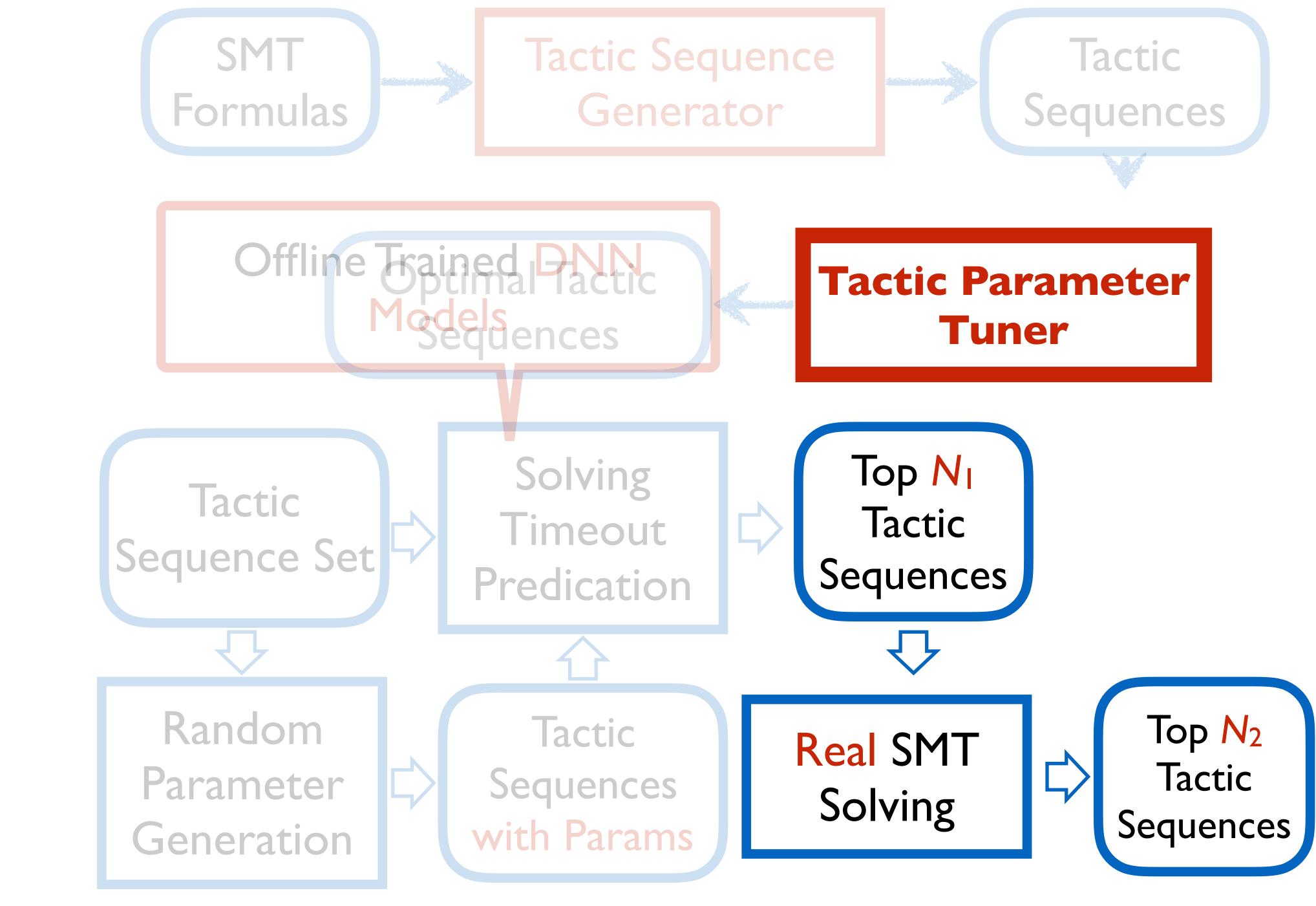
any parameter

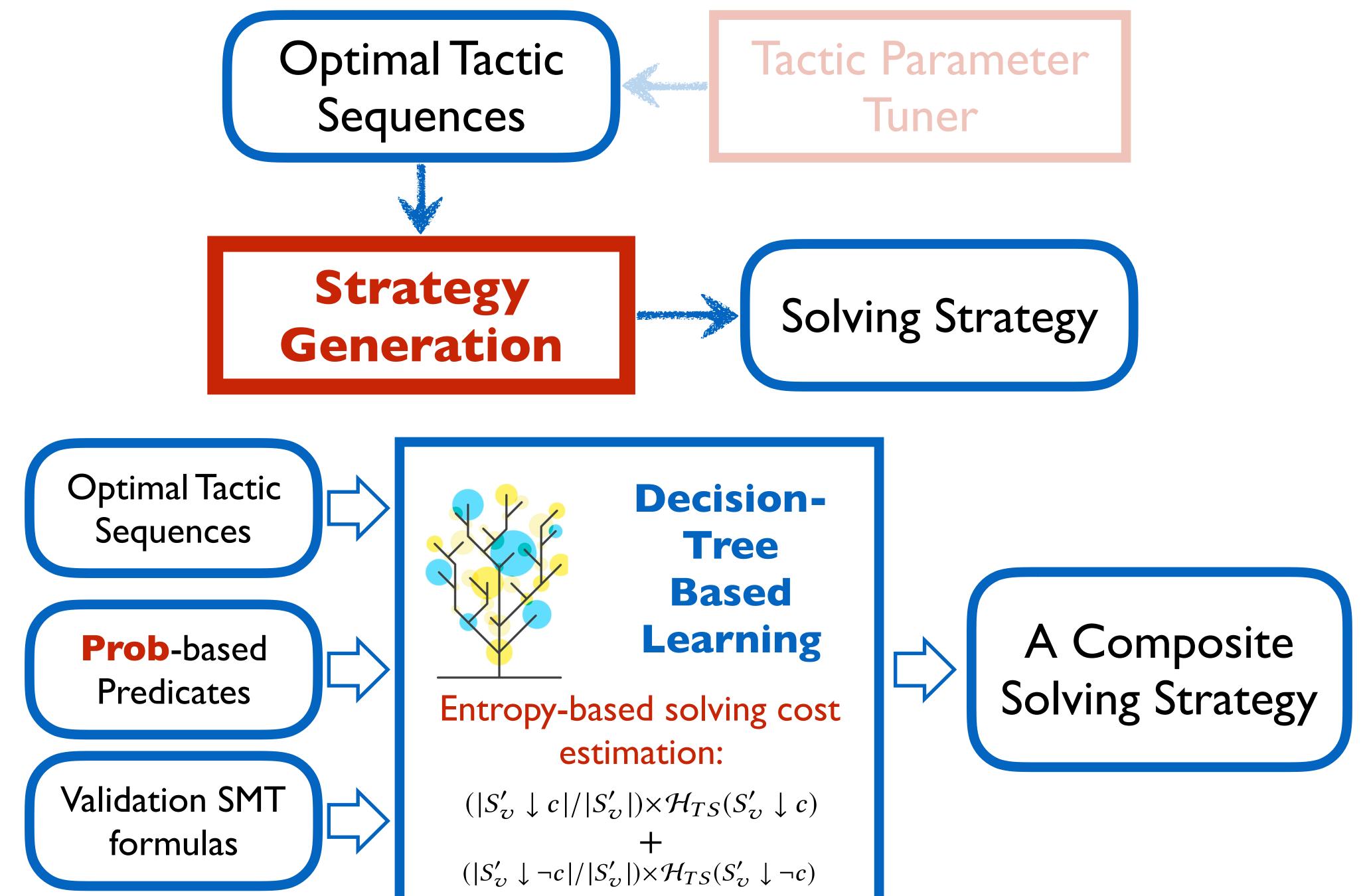












- Research questions
  - Effectiveness
    - Queries (solved formulas)
    - Paths
  - Generalization ability

- Implementation
  - KLEE with Z3
  - Pytorch for ML models
  - JPF-based concolic engine

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  - KLEE with Z3
  - Pytorch for ML models
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The deep learning models trained for C programs are directly used for Java programs

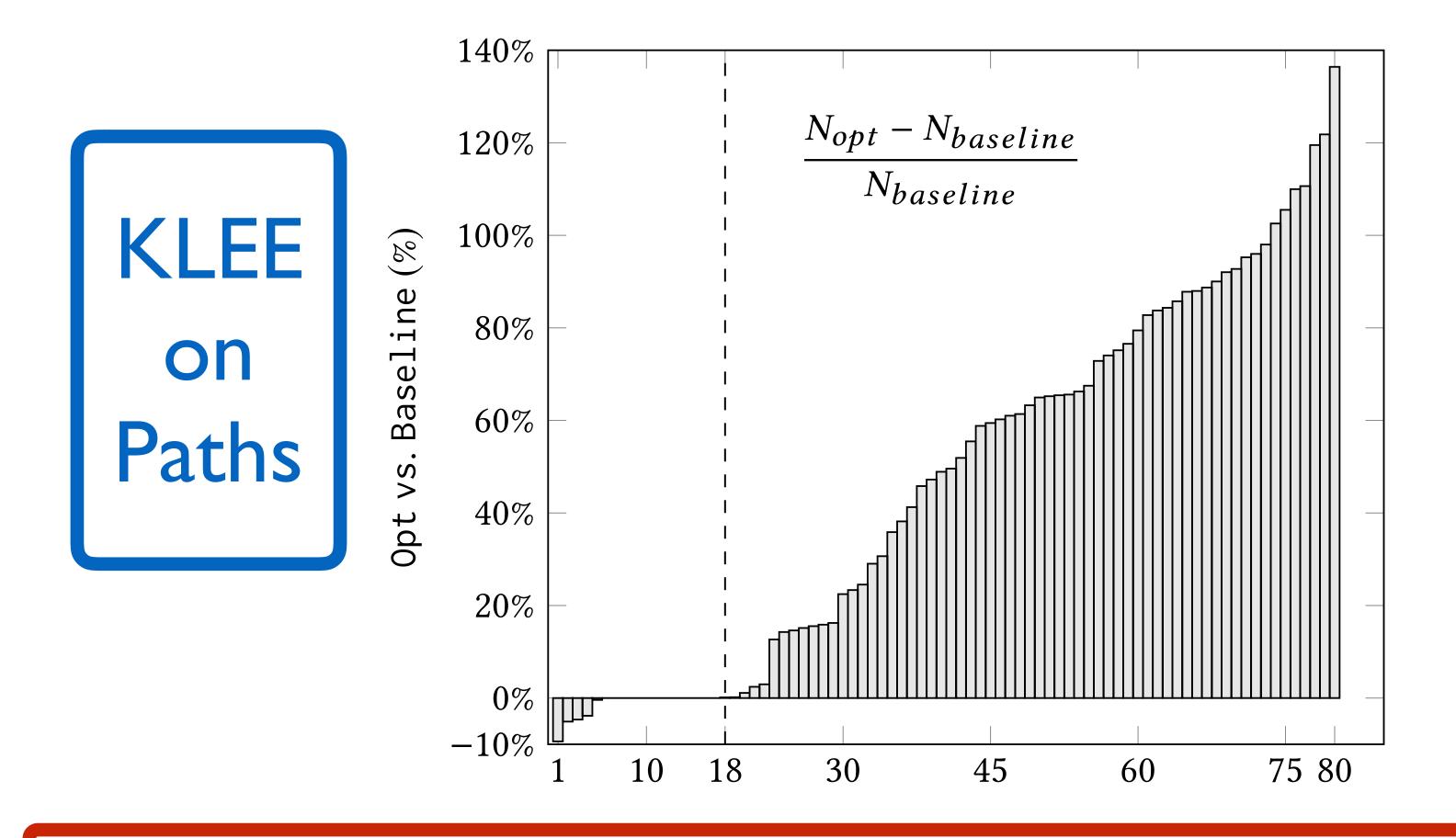
# ML Models Training

- DRL model for generating tactic sequences
  - I4 randomly selected Coreutils programs
  - 300 SMT formulas randomly from each
  - 4200 in total

# ML Model Training

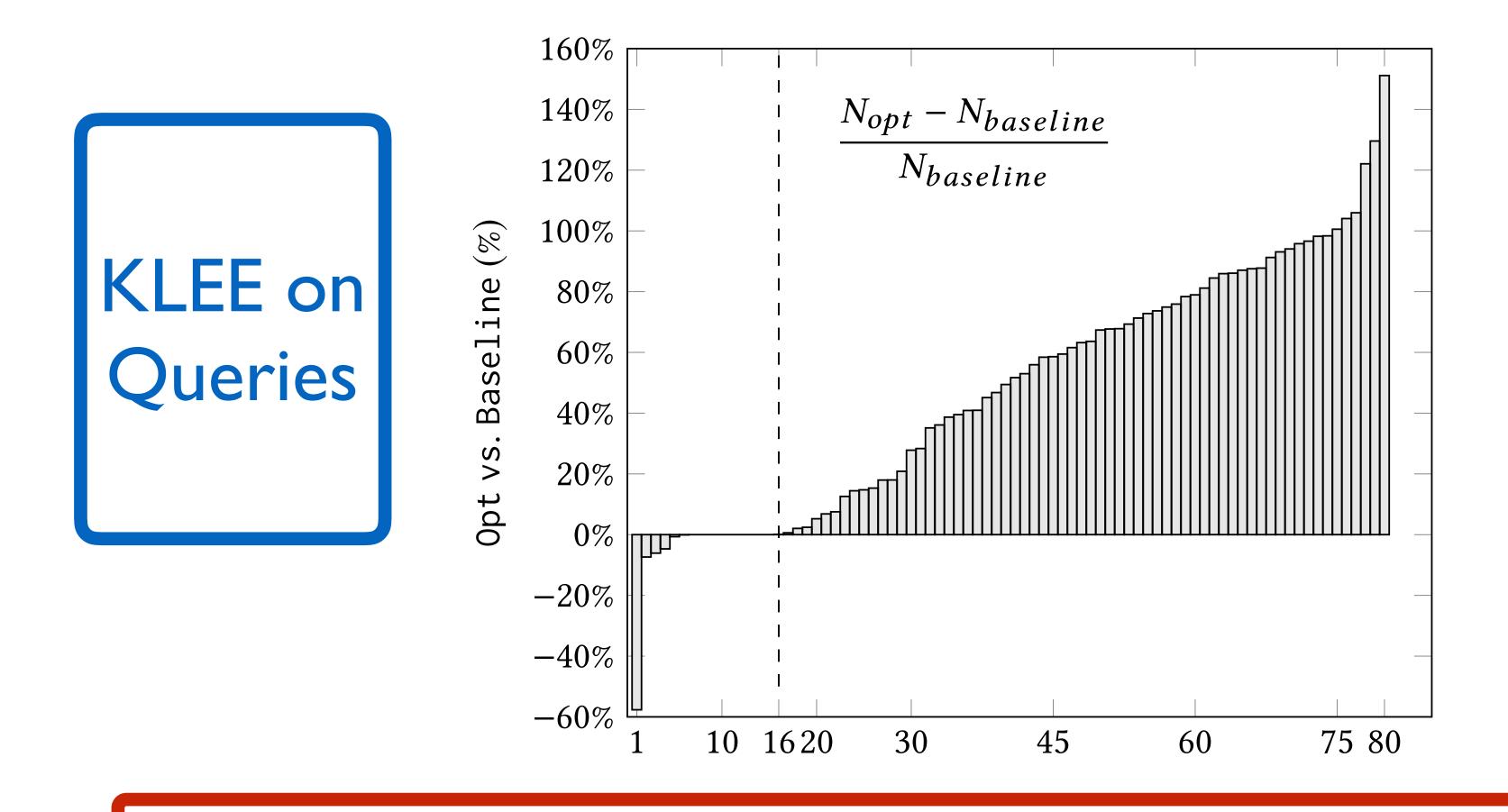
- DNN models for predicating timeout
  - SAT, SMT, QFNRA-NLSAT, QFNRA
  - 8 randomly selected Coreutils programs
  - QF BV, QF ABV, QF ABVFP, QF BVFP SMT-LIB2 benchmarks • Timeout threshold: 30 seconds

- Benchmark for KLEE
  - 80 Coreutils programs
- Benchmark for Java concolic engine
  - 34 open-source Java programs
  - 327506 SLOCs in total



### Results of Effectiveness

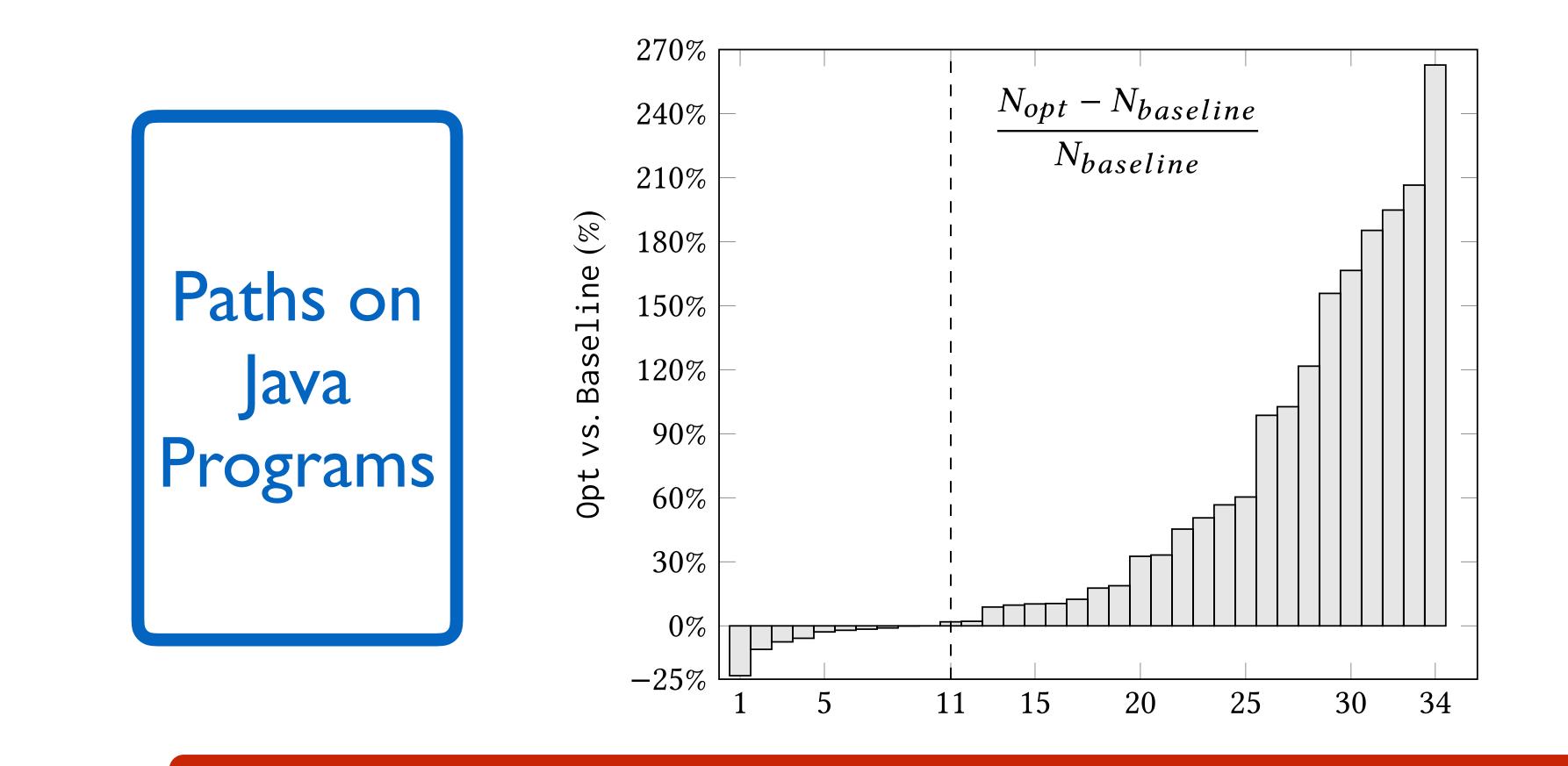
Improve the explored paths for 63 programs, 66.11% on average



### Results of Effectiveness

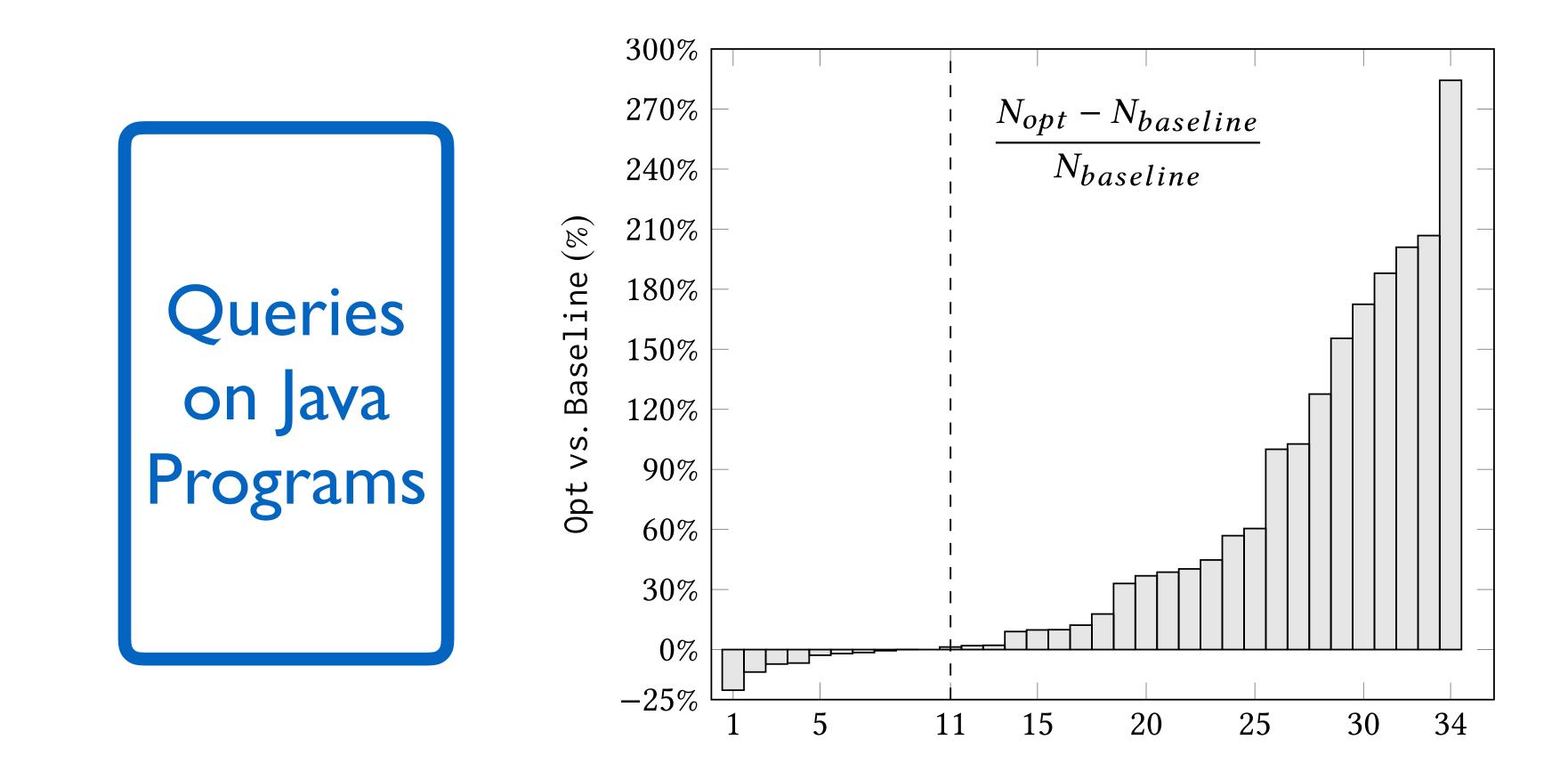
Improve the queries for 65 programs, 58.76% on average

### Results of Generalization Ability



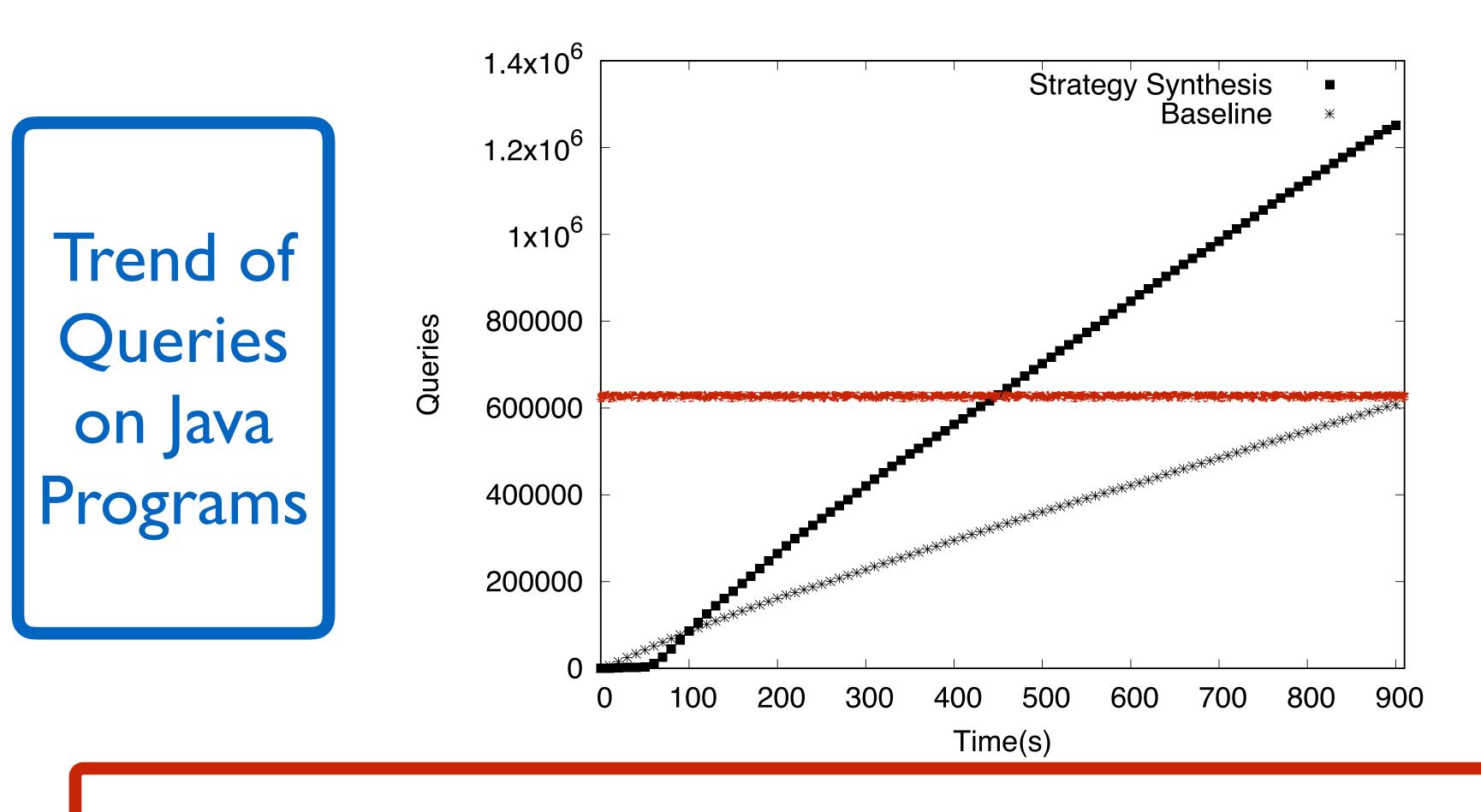
Improve the explored paths for 24 programs, 102.6% on average

### Results of Generalization Ability



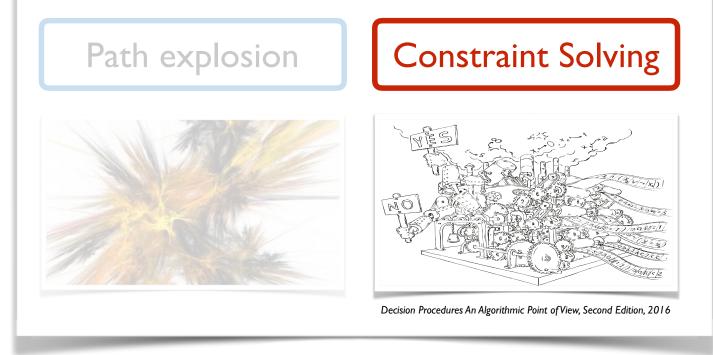
Improve the queries for 24 programs, 100.24% on average

### Results of Generalization Ability

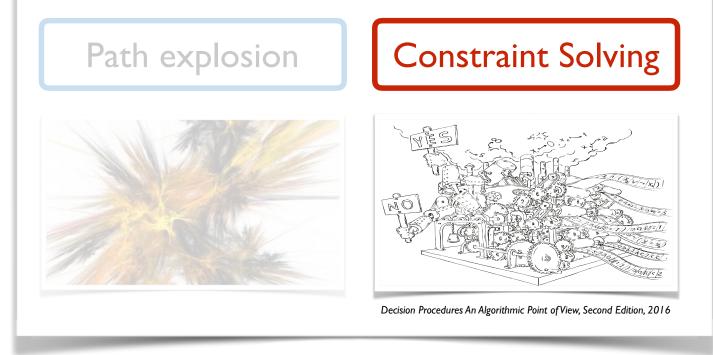


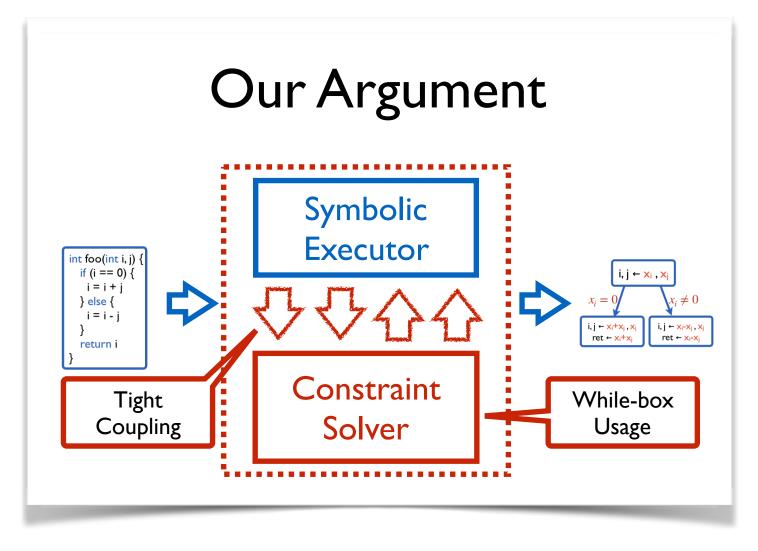
### 2.07x speedup for solving the same amount of queries

### Our Work's Target

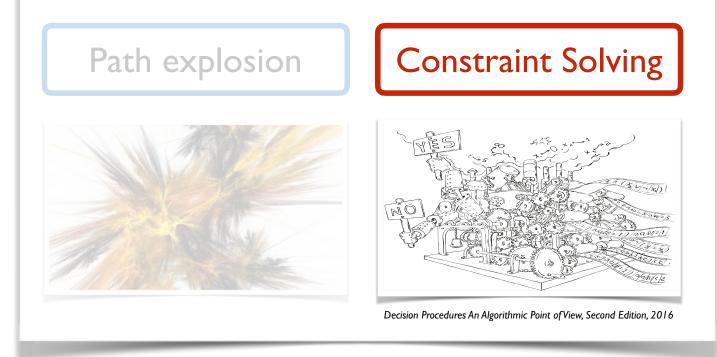


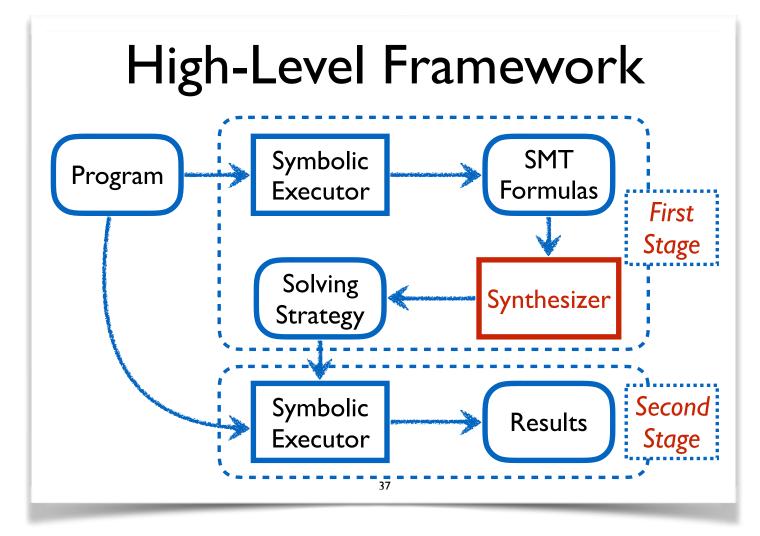
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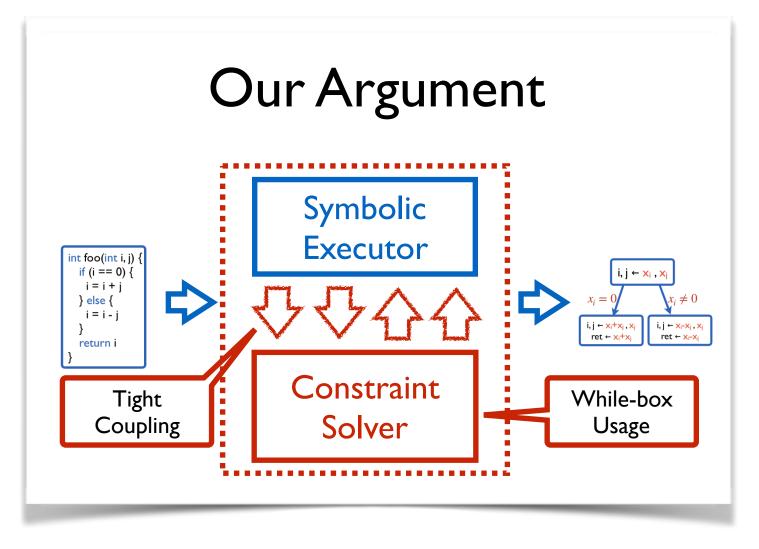




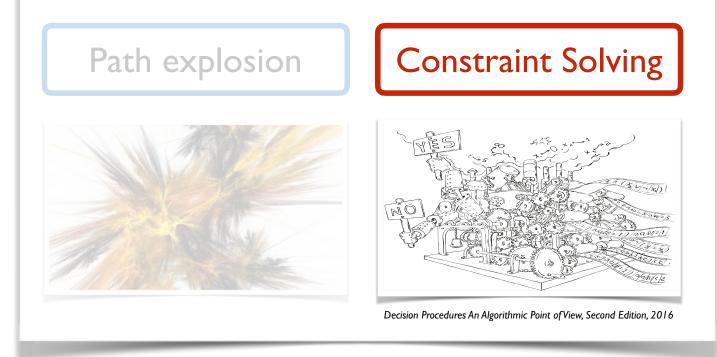
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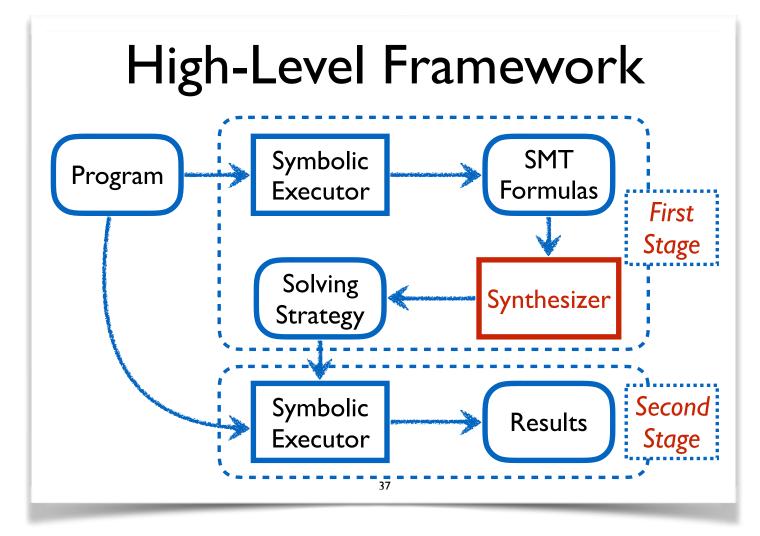


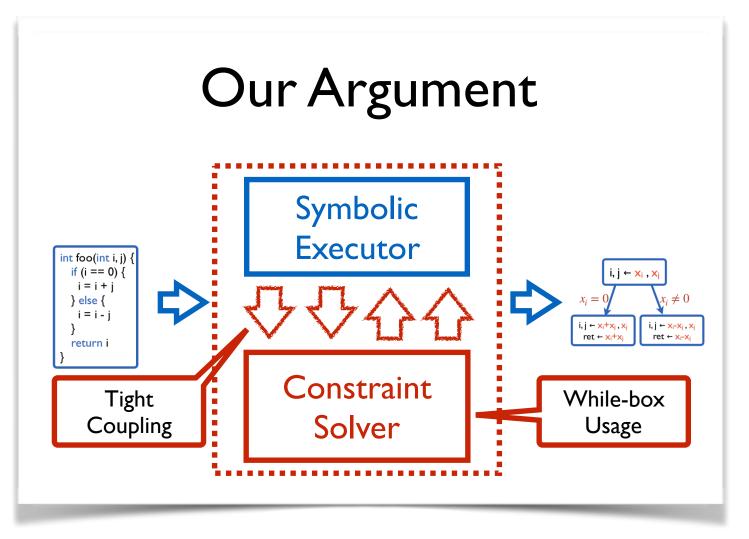


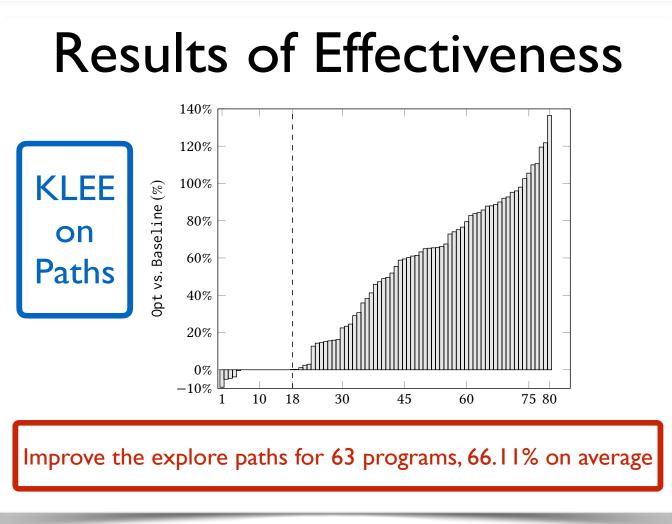


### Our Work's Target









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## Thank you! Q&A

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