

Syntax-Guided Program Synthesis

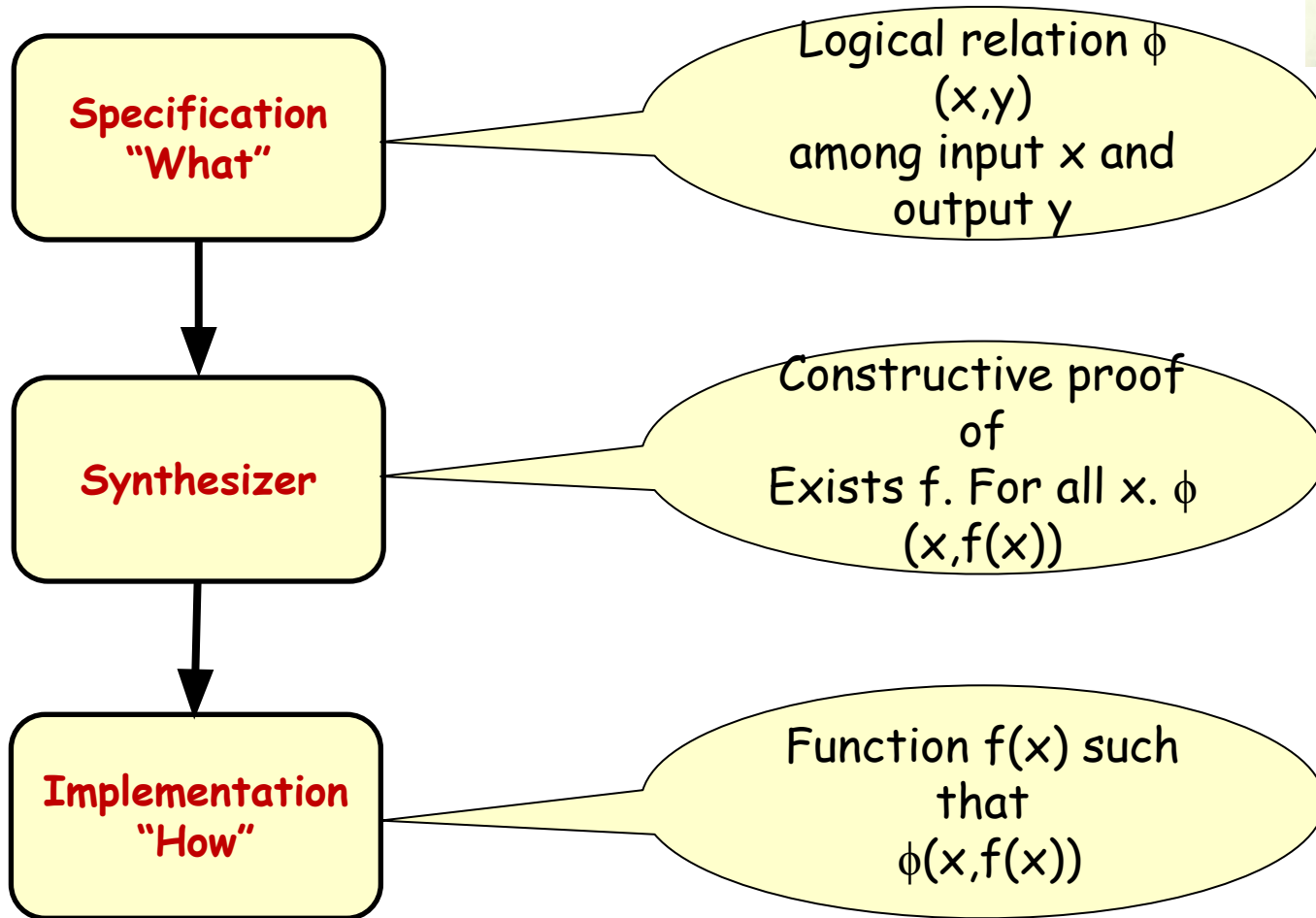
Rajeev Alur

Workshop on Program Synthesis
for Scientific Computing, August 2020

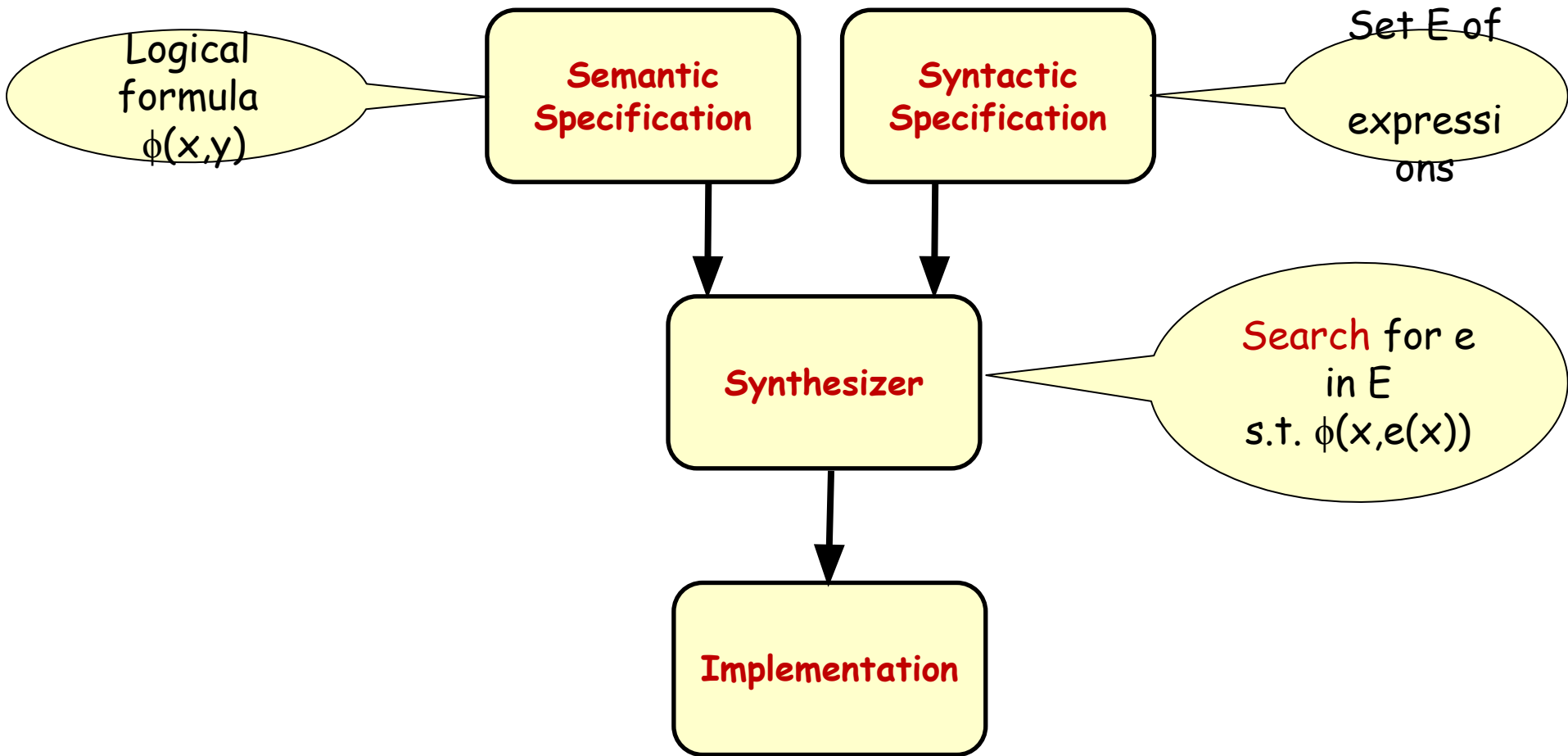


Classical Program Synthesis

Church (1957)



Syntax-Guided Search-based Program Synthesis



Motivating Applications

- ❑ Superoptimizing compilers: Given a program fragment P , find a functionally equivalent program with resource constraints (e.g. fewer instructions, or avoid certain expensive instructions)
- ❑ Program repair: Automatically edit a program locally to fix a bug (particularly helpful to students in Intro Programming courses)
- ❑ Proof objects for verification: template-guided synthesis of inductive invariants, ranking functions, program analysis rules, ...
- ❑ Programming by examples / demonstration: Can non-programmers communicate intent intuitively?

Syntax-Guided Synthesis (SyGuS)



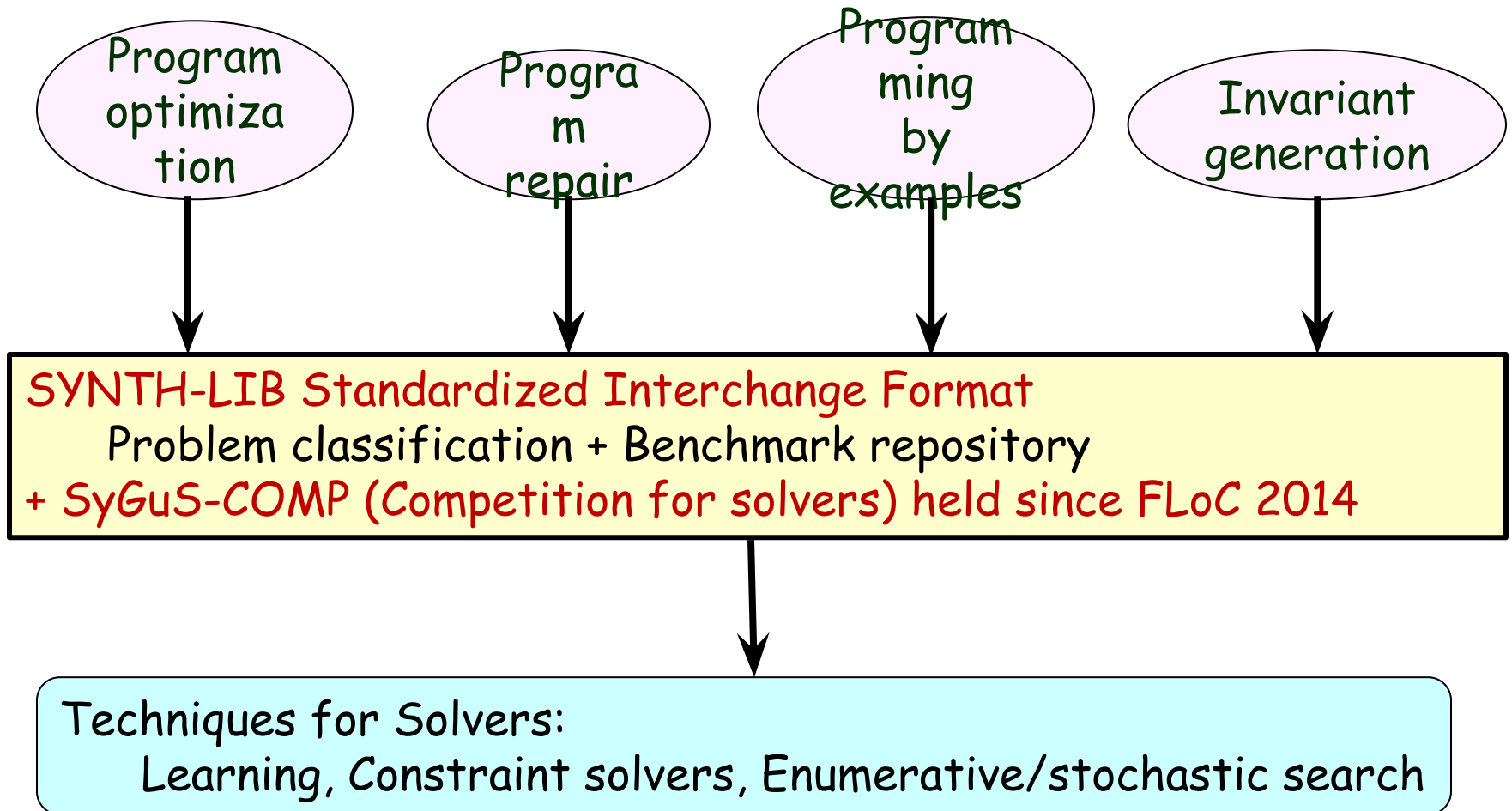
www.syguS.org

- Fix a background theory T : fixes types and operations
- Function to be synthesized: name f along with its type
 - General case: multiple functions to be synthesized
- Inputs to SyGuS problem:
 - Specification $\phi(x, f(x))$
Typed formula using symbols in $T + \text{symbol } f$
 - Set E of expressions given by a context-free grammar
Set of candidate expressions that use symbols in T
- Computational problem:
Output e in E such that $\phi[f/e]$ is valid (in theory T)

Syntax-guided synthesis; FMCAD'13

with Bodik, Juniwal, Martin, Raghothaman, Seshia, Singh, Solar-Lezama, Torlak, Udupa

SyGuS Competition



SyGuS Progress



www.syguS.org

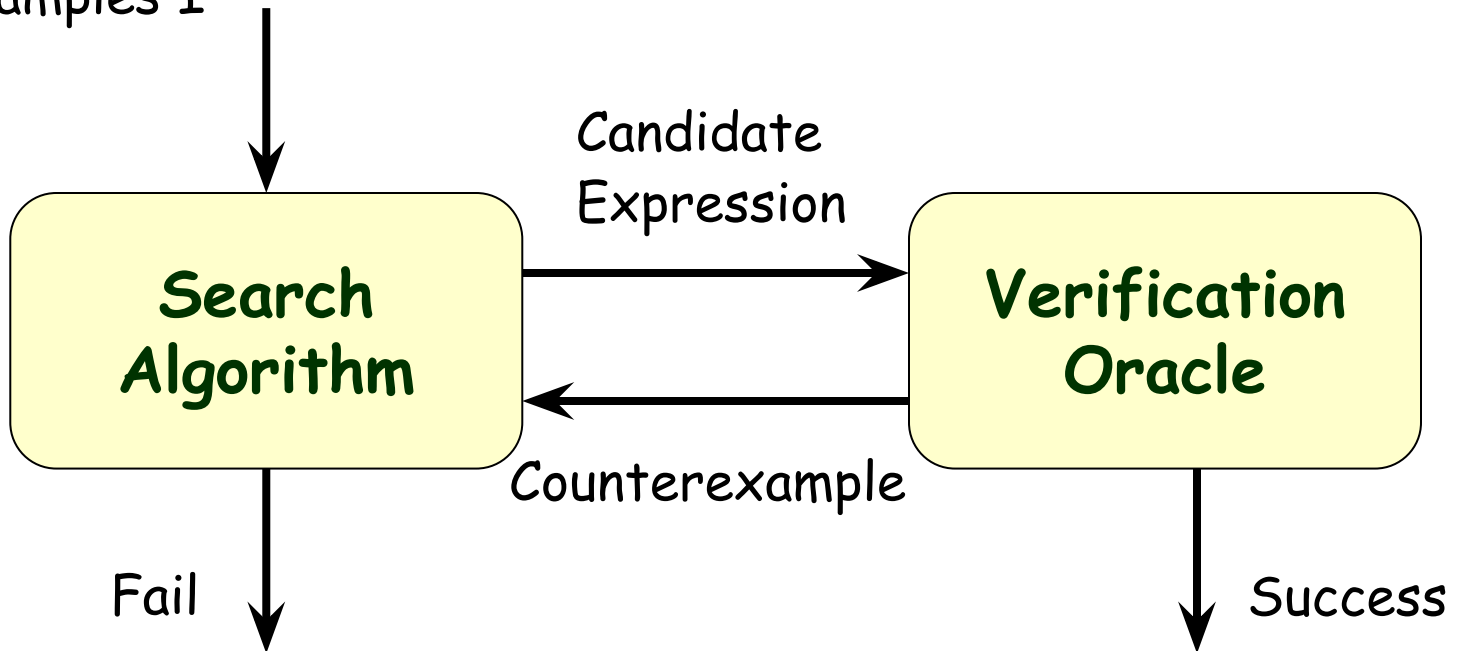
- ❑ Over 2800 benchmarks
 - Hacker's delight
 - Invariant generation (based on verification competition SV-Comp)
 - FlashFill (programming by examples system from Microsoft)
 - Synthesis of attack-resilient crypto circuits
 - Program repair
 - Motion planning
 - ICFP programming competition

- ❑ Special tracks for competition
 - Invariant generation
 - Programming by examples
 - Conditional linear arithmetic

- ❑ Current winner: CVC4 (Reynolds et al) search integrated in constraint solving

Search and Verify

Initial examples I



Concept class: Set E of expressions

Examples: Concrete input values

Counterexample-guided Inductive Synthesis (CEGIS)

Goal: Find f such that for all x in D , $\phi(x, f)$ holds

$I = \{ \};$ /* Interesting set of inputs */

Repeat

Learn: Find f such that for all x in I , $\phi(f, x)$ holds

Verify: Check if for all x in D , $\phi(f, x)$ holds

If so, return f

If not, find x such that $\sim \phi(f, x)$ holds, and add x to I

Implementing Search

□ Given:

Specification $\phi(x, f(x))$

Grammar for set E of candidate implementations

Finite set I of inputs

Find an expression $e(x)$ in E s.t. $\phi(x, e(x))$ holds for all x in I

- Enumerative search with lots of optimizations for pruning
- Symbolic constraints over variables encoding desired expression tree
- Stochastic search in spirit of genetic programming
- Divide and conquer strategies to build sub-expressions
- Partial evaluation to rule candidates before fully expanding them
- Establishing unrealizability of synthesis
- Type-directed enumeration

Acceleration Using Learned Probabilistic Models

- Can we bias the search towards likely programs?
- Step 1: Mine existing solutions to convert given grammar into a probabilistic higher-order grammar
 - Weighted production rules
 - Conditioned on parent and sibling context
 - Transfer learning used to avoid overfitting
- Step 2: Enumerative search to generate expressions in decreasing likelihood
 - Use A^* with cost estimation heuristic
 - Integrated with previous optimizations (equivalence-based pruning...)

With W. Lee, K. Heo, and M. Naik (PLDI 2018)

Future Directions

- ❑ Beyond SMT Solvers: SyGuS-like back-end focused on efficient search, but decoupled from SMT solvers so as to allow interface with alternative testing / verification tools
- ❑ More theories, benchmarks, and applications: tables and relational queries, floating point arithmetic
- ❑ Quantitative synthesis and optimization
- ❑ Applications in scientific computing ??