# Synthesizing Programs with Constraint Solvers 

ASPLOS Symposium

Ras Bodik Division of Computer Science University of California, Berkeley

## Prepare your language for synthesis

Extend the language with two constructs

result: int bar (int x) implements foo \{ return $x$ << 1;
\}
instead of implements, assertions over safety properties can be used

## Synthesis as search over candidate programs

Partial program (sketch) defines a candidate space we search this space for a program that meets $\phi$

Usually can't search this space by enumeration space too large (>> $10^{10}$ )

Describe the space symbolically
solution to constraints encoded in a logical formula gives values of holes, indirectly identifying a correct program

What constraints? Essentially encode semantics in SAT

## Synthesis from partial programs



## Example: Parallel Matrix Transpose

## Example: 4×4-matrix transpose with SIMD

## a functional (executable) specification:

```
int[16] transpose(int[16] M) \{
    int[16] T = 0;
    for (int i = 0; i < 4; i++)
        for (int \(\mathrm{j}=0\); j < 4; j++)
        \(\mathrm{T}[4\) * \(\mathrm{i}+\mathrm{j}]=\mathrm{M}[4\) * \(\mathrm{j}+\mathrm{i}]\);
    return T ;
\}
```

This example comes from a Sketch grad-student contest

## Implementation idea: parallelize with SIMD

## Intel SHUFP (shuffle parallel scalars) SIMD instruction:

$$
\text { return }=\text { shufps(x1, x2, imm8 :: bitvector8) }
$$



## High-level insight of the algorithm designer

Matrix $M$ transposed in two shuffle phases

Phase 1: shuffle $M$ into an intermediate matrix $S$ with some number of shufps instructions

Phase 2: shuffle $S$ into an result matrix $T$ with some number of shufps instructions

Synthesis with partial programs helps one to complete their insight. Or prove it wrong.

## The SIMD matrix transpose, sketched

```
int[16] trans_sse(int[16] M) implements trans {
    int[16] S = 0, T = 0;
    S[??::4] = shufps(M[??::4], M[??::4], ??);
    S[??::4] = shufps(M[??::4], M[??::4], ??);
                            Phase 1
S[??::4] = shufps(M[??::4], M[??::4], ??); ]
T[??::4] = shufps(S[??::4], S[??::4], ??);
T[??::4] = shufps(S[??::4], S[??::4], ??);
                                    Phase 2
T[??::4] = shufps(S[??::4], S[??::4], ??);
    return T;

\section*{The SIMD matrix transpose, sketched}
```

int[16] trans_sse(int[16] M) implements trans {
int[16] S = 0, T = 0;
repeat (??) S[??::4] = shufps(M[??::4], M[??::4], ??);
repeat (??) T[??::4] = shufps(S[??::4], S[??::4], ??);
return T;
}
int[16] trans_sse(int[16] M) implements trans { // synthesized code
S[4::4] = shufps(M[6::4], M[2::4], 11001000b);
S[0::4] = shufps(M[11::4], M[6::4], 10010110b);
S[12::4] = shufps(M[0::4], M[2::4], 10001101b);
S[8::4] = shufps(M[8::4], M[12::4], 11010111b);
T[4::4] = shufps(S[11::4], S[1::41, 10111100b):
T[12::4] = shufps(S[3 From the contestant email:
T[8::4] = shufps(S[4 Over the summer, I spent about 1/2
T[0::4] = shufps(S[1 a day manually figuring it out.
Synthesis time: <5 minutes.

```

\section*{Demo: transpose on Sketch}

\section*{Try Sketch online at http://bit.ly/sketch-language}

\section*{Demo notes (1)}

In the demo, we accelerated synthesis by changing
repeat(??) loop body
repeat(??) loop body
to
int steps = ??
repeat(steps) loop body
repeat(steps) loop body
\(\rightarrow\) can improve efficiency by adding more "insight" here, the "insight" constraints state that both loops have same (unknown) number of iterations

\section*{Demo notes (2)}

How did the student come up with the insight that two phases are sufficient?

We don't know but the synthesizer can prove that one phase is insufficient (a one-phase sketch has no solution)

\section*{Program Synthesis with Constraint Solvers}

\section*{What to do with a program as a formula?}

Assume a formula \(\mathrm{S}_{\mathrm{P}}(\mathrm{x}, \mathrm{y})\) which holds iff program \(\mathrm{P}(\mathrm{x})\) outputs value y
```

program: f(x) { return x + x }

```
formula: \(\quad S_{f}(x, y): y=x+x\)

This formula is created as in program verification with concrete semantics [CMBC, Java Pathfinder, ...]

\section*{With program as a formula, solver is versatile}

Solver as an interpreter: given \(x\), evaluate \(f(x)\)
\[
S(x, y) \wedge x=3 \quad \text { solve for } y \quad \boldsymbol{y} \mapsto \mathbf{6}
\]

Solver as a program inverter: given \(f(x)\), find \(x\)
\[
S(x, y) \wedge y=6 \quad \text { solve for } x \quad \boldsymbol{x} \mapsto \mathbf{3}
\]

This solver "bidirectionality" enables synthesis

\section*{Search of candidates as constraint solving}
\(S_{P}(x, h, y)\) holds iff sketch \(P[h](x)\) outputs \(y\). spec(x) \{ return \(\mathrm{x}+\mathrm{x}\}\)
sketch \((\mathrm{x})\) \{ return x << ?? \} \(\quad S_{\text {sketch }}(x, y, h): y=x * 2^{h}\)
The solver computes \(h\), thus synthesizing a program correct for the given x (here, \(\mathrm{x}=2\) )
\[
S_{\text {sketch }}(x, y, h) \wedge x=2 \wedge y=4 \quad \text { solve for } h \quad \boldsymbol{h} \mapsto \mathbf{1}
\]

Sometimes \(h\) must be constrained on several inputs
\[
\begin{aligned}
& S\left(x_{1}, y_{1}, h\right) \wedge x_{1}=0 \wedge y_{1}=0 \wedge \\
& S\left(x_{2}, y_{2}, h\right) \wedge x_{2}=3 \wedge y_{2}=6 \quad \text { solve for } h \quad \boldsymbol{h} \mapsto \mathbf{1}
\end{aligned}
\]

\section*{Inductive synthesis}

Our constraints encode inductive synthesis:
We ask for a program \(P\) correct on a few inputs. We hope (or test, verify) that \(P\) is correct on rest of inputs.

How to select suitable inputs?
Verify a candidate program. If it fails verification, the counterexample (input) is added as an input to synthesis

\section*{More information}

\section*{Learn:}
- CAV 2012 invited tutorial (with Emina Torlak)
- graduate seminar (cs294-fa12)

Play:
- SKETCH synthesizer
- Rosette lightweight synthesizer

\section*{Acknowledgements}

\section*{UC Berkeley}

Gilad Arnold
Shaon Barman
Prof. Ras Bodik
Prof. Bob Brayton
Joel Galenson
Thibaud Hottelier
Sagar Jain
Chris Jones
Ali Sinan Koksal
Leo Meyerovich
Evan Pu

\section*{MIT}

Casey Rodarmor
Prof. Koushik Sen
Prof. Sanjit Seshia
Lexin Shan
Saurabh Srivastava
Liviu Tancau
Nicholas Tung
IBM
Satish Chandra
Kemal Ebcioglu
Rodric Rabbah
Vijay Saraswat
Vivek Sarkar```

