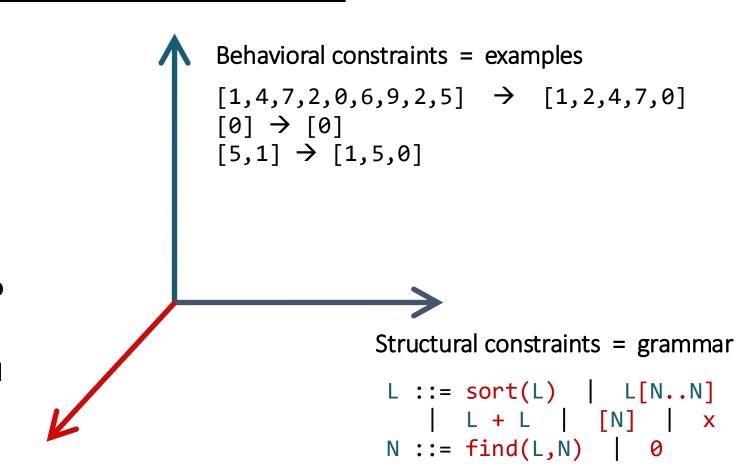
Lecture 6 Stochastic Search

The problem statement



Search strategy?

Enumerative Representation-based **Stochastic**

Constraint-based

Stochastic search in synthesis

Weimer, Nguyen, Le Goues, Forrest. *Automatically Finding Patches Using Genetic Programming*. ICSE'09

Gissurarson, Applis, Panichella, van Deursen, Sands. *PropR: Property-Based Automatic Program Repair.* ICSE'22

Schkufza, Sharma, Aiken: *Stochastic superoptimization*. ASPLOS 2013

Shi, Steinhardt, Liang: FrAngel: Component-Based Synthesis with Control Structures. POPL'19

Stochastic search in synthesis

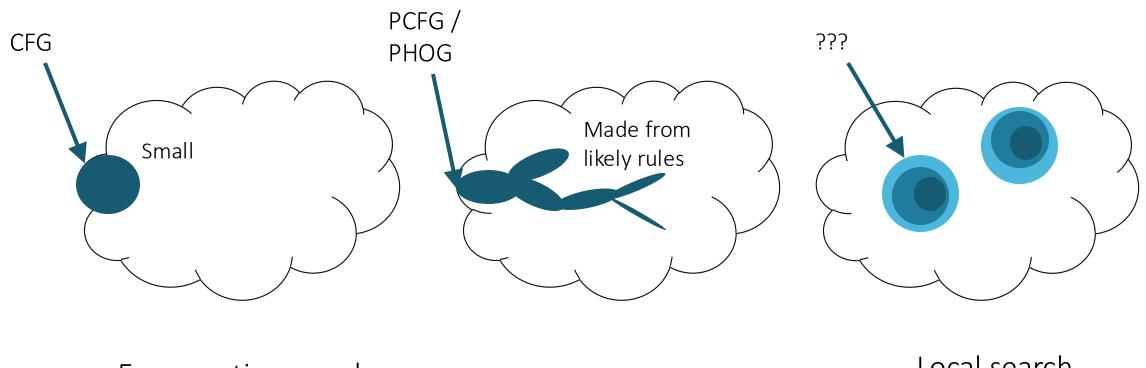
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Search space



Enumerative search

Weighted enumerative search

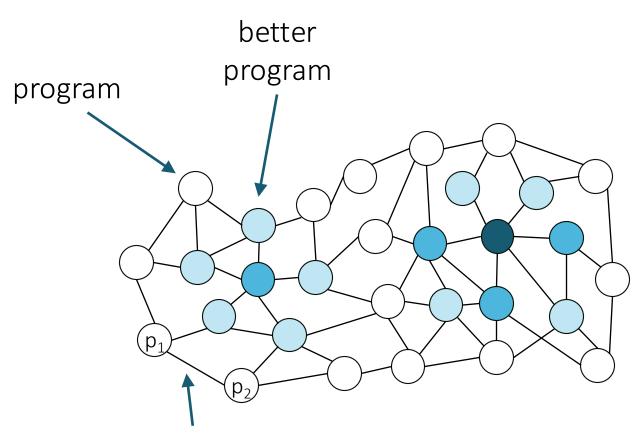
Local search

Naïve local search

To find the best program:

```
p := random()
while (true) {
   p' := mutate(p);
   if (C(p') < C(p))
      p := p';
}</pre>
```

Will never get to \bigcirc from $p_1!$



can generate p_2 from p_1 (and vice versa) via mutation

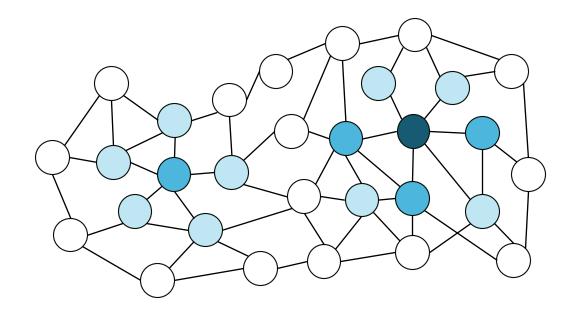
MCMC sampling

Avoid getting stuck in local minima:

```
p := random()
while (true) {
   p' := mutate(p);
   if (random(A(p -> p'))
      p := p';
}
```

where

- if p' is better than p: $A(p \rightarrow p') = 1$
- otherswise: $A(p \rightarrow p')$ decreases with difference in cost between p' and p



MCMC sampling

Metropolis algorithm:

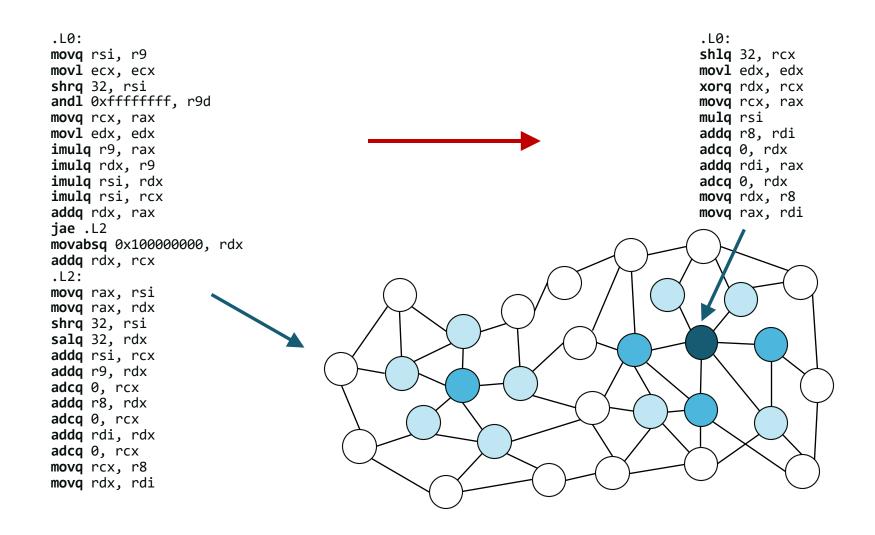
$$A(p \to p') = \min(1, e^{-\beta(C(p') - C(p))})$$

The theory of Markov chains tells us that in the limit we will be sampling with the probability proportional to

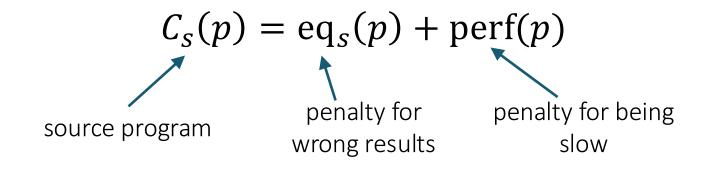
$$e^{-\beta*C(p)}$$

MCMC for superoptimization

[Schkufza, Sharma, Aiken '13]



Cost function



$$\operatorname{eq}_{s}(p) = \sum_{t \in Tests} \operatorname{reg}_{s}(p, t) + \operatorname{mem}_{s}(p, t) + \operatorname{err}(p, t)$$

$$\uparrow$$

$$\text{# of different bits in} \qquad \text{# of segfaults etc}$$

$$\operatorname{registers/memory}$$

when $eq_s(p) = 0$, use a symbolic validator (to check actual equivalence)

Cost function

$$C_s(p) = \operatorname{eq}_s(p) + \operatorname{perf}(p)$$
source program

penalty for penalty for being wrong results slow

$$perf(p) = \sum_{i \in instr(p)} latency(i)$$

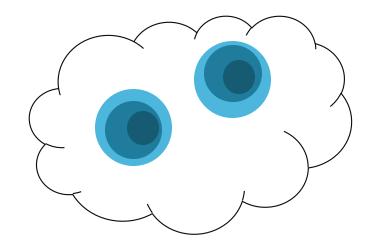
Local search: discussion

Strengths:

• can explore program spaces with no a-priori bias

Limitations?

- only applicable when there is a cost function that faithfully approximates correctness
- Counterexample: round to next power of two
 - 0011 -> 0100
 - 0101 -> 1000
 - 0111 -> 1000



Stochastic search in synthesis

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Gissurarson, Applis, Panichella, van Deursen, Sands. *PropR: Property-Based Automatic Program Repair.* ICSE'22

• Similar but for program repair, uses genetic programming

Schkufza, Sharma, Aiken: *Stochastic superoptimization*. ASPLOS 2013

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- Samples from a grammar with bias towards partial solutions
- I assume they use stochastic just for ease of sampling

Next



Enumerative Representation-based Stochastic

Constraint-based

