#### **#11: Stochastic Search**

#### Sankha Narayan Guria

EECS 700: Introduction to Program Synthesis



## The problem statement



 $[6] \rightarrow [6]$  $[5,1] \rightarrow [1,5,0]$ 

#### **Search strategy?**

Enumerative
Representation-based **Stochastic**Constraint-based

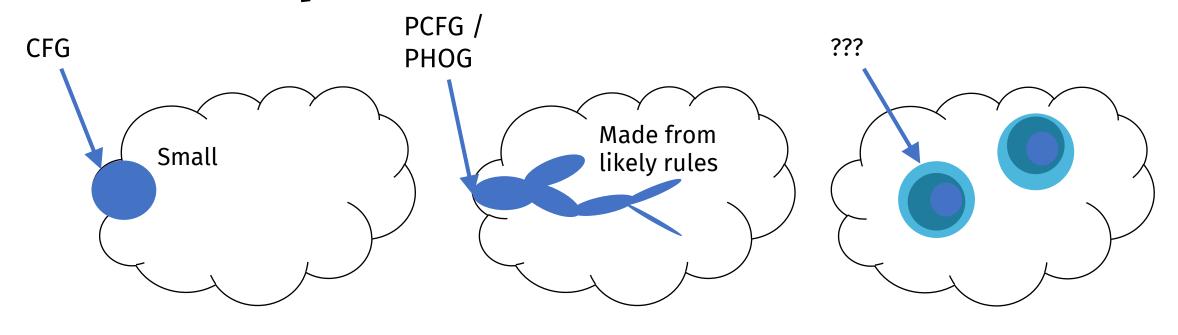
## Stochastic search in synthesis

- 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'13
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### **Search space**



Enumerative search

Weighted enumerative search

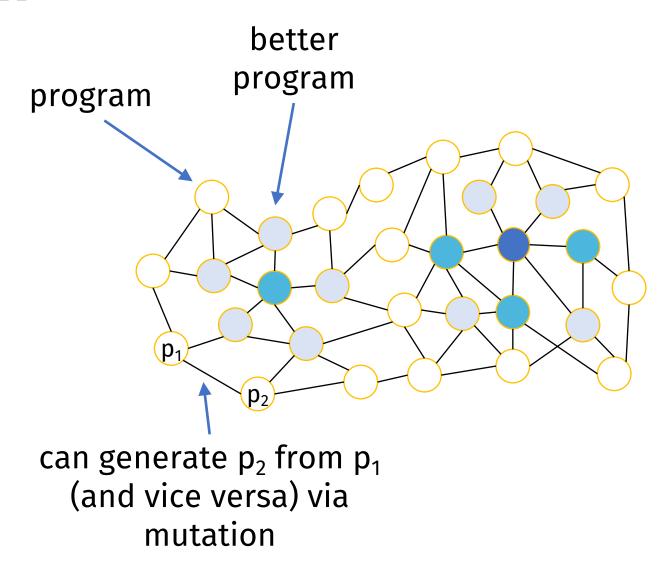
Local search

#### Naïve local search

• To find the best program:

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

• Will never get to  $\bigcirc$  from  $p_1!$ 



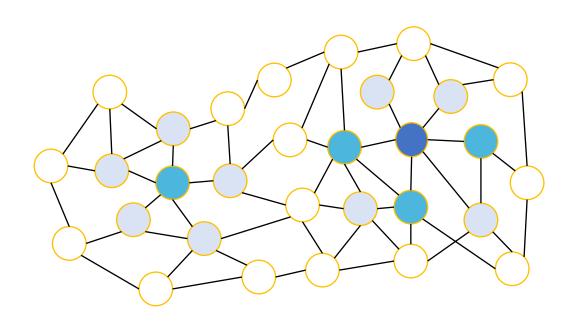
# MCMC sampling

 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:  $A(p \rightarrow p') = 1$
- otherwise:  $A(p \rightarrow p')$  decreases with difference in cost between p' and p



# MCMC sampling

Metropolis algorithm:

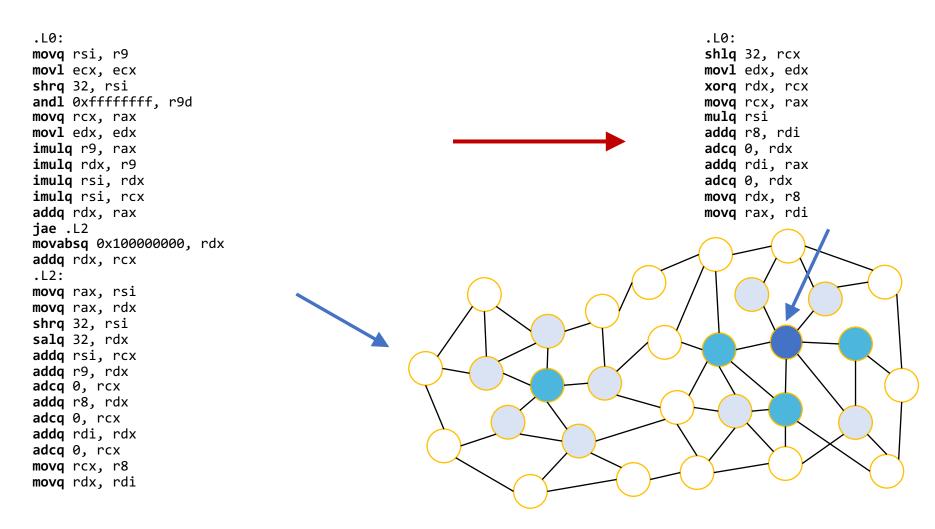
$$A(p \to p') = \min(1, e^{-\beta(C(p') - C(p))})$$

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

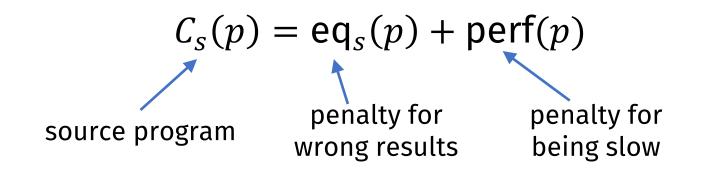
$$e^{-\beta *C(p)}$$

## MCMC for superoptimization

[Schkufza, Sharma, Aiken '13]



#### **Cost function**



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

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

$$\operatorname{in}$$

$$\operatorname{registers/memory}$$

$$\operatorname{when } \operatorname{eq}_s(p) = 0, \text{ use a symbolic validator}$$

#### **Cost function**

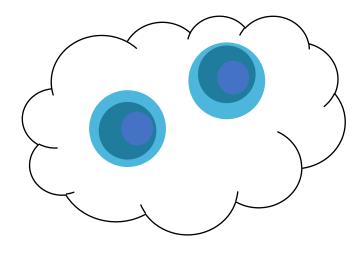
$$C_S(p) = \operatorname{eq}_S(p) + \operatorname{perf}(p)$$
source program

penalty for penalty for wrong results being 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



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  - Similar but for program repair, uses genetic programming
- Schkufza, Sharma, Aiken: Stochastic superoptimization. ASPLOS'13
- Shi, Steinhardt, Liang: FrAngel: Component-Based Synthesis with Control Structures. POPL'19
  - Samples from a grammar with bias towards partial solutions