

# Efficient Simulation Of A Simple Evolutionary System

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# Outline

Background

Question 1: Distance between finite and infinite population

Question 2: Oscillation in finite population

Question 3: Oscillation in finite population under violation in mutation

Question 4: Oscillation in finite population under violation in crossover

Conclusion

# Terms

Population  $P$ : a collection of length  $\ell$  binary strings

Population vector  $\mathbf{p}$ :  $\mathbf{p}_j$  is the proportion of string  $j$  in the population

If  $P = 00, 01, 01, 10, 11, 11$ , then  $\mathbf{p}_3 = 2/6 = 1/3$

$\mathcal{R}$  denotes a set binary strings of length  $\ell$

Addition and multiplication of elements in  $\mathcal{R}$  are bitwise operations modulo 2

$$x = 1101, y = 1010$$

$$x + y = 1101 + 1010 = 0111$$

$$xy = 1101 \cdot 1010 = 1000$$

$$\bar{x} = 0010$$

# Crossover & Mutation

Crossover : Choose parents  $u$  and  $v$ , exchange bits using crossover mask  $m$ :

$$u' = um + v\bar{m}, v' = u\bar{m} + vm$$

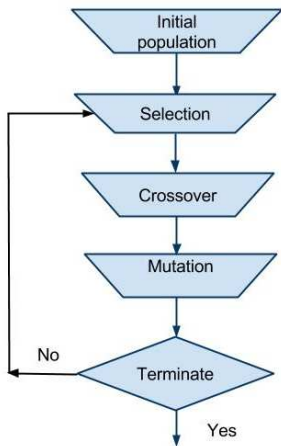
$$u = \mathbf{11001011}, v = 11011111, m = 11110000$$

$$\{\mathbf{11001011}, 11011111\} \rightarrow \{\mathbf{11000000} + 00001111, 0000\mathbf{1011} + 11010000\} \rightarrow \{\mathbf{11001111}, 1101\mathbf{1011}\}$$

Mutation: Flip bits using mutation mask:

$$x \rightarrow x + m$$

# Finite Population GA



Randomly select parents  $u$  and  $v$

Crossover  $u$  and  $v$  to produce  $u'$  and  $v'$

Keep one of  $u'$ ,  $v'$ , and mutate

Repeat above to form next generation

Repeat whole process until system stops to improve or threshold is reached

# Infinite Population Model

Population is modeled as a vector  $\mathbf{p}$

$\mathcal{G}$  maps  $\mathbf{p}$  to the next generation

$\mathcal{G}(\mathbf{p})_j$  = probability that string  $j$  occurs in the next generation

The infinite population model is the sequence

$$\mathbf{p} \rightarrow \mathcal{G}(\mathbf{p}) \rightarrow \mathcal{G}(\mathcal{G}(\mathbf{p})) \rightarrow \dots$$

# Random Heuristic Search

$\tau$  is a stochastic transition rule that maps  $\mathbf{p}$  to  $\mathbf{p}$

For a finite population, sequence  $\mathbf{p}, \tau(\mathbf{p}), \tau^2(\mathbf{p}), \dots$  forms Markov chain

$\tau(\mathbf{p})$  cannot be predicted with certainty

$\mathcal{G}(\mathbf{p})$  is the expected next generation  $\mathcal{E}(\tau(\mathbf{p}))$

The variance in the next generation is

$$\mathcal{E}(\|\tau(\mathbf{p}) - \mathcal{G}(\mathbf{p})\|^2) = \frac{1 - \|\mathcal{G}(\mathbf{p})\|^2}{r}$$

# History

Haldane, in 1932, summarized basic population genetics models : Wright, Fisher and Haldane

Several people working with evolution-inspired algorithms in the 1950s and the 1960s Box (1957), Friedman(1959), Bledsoe (1961), Bremermann (1962), and Reed, Toombs and Baricelli (1967)

In 1960s and 1970s, Holland and colleagues formalized and promoted population based algorithms with crossover and mutation

Vose (1999) presented efficient methods for computing with a haploid model using mask-based operators introduced by Geiringer (1944)