#### CEng 713, Evolutionary Computation, Lecture Notes



### Schema Theorem

- Holland's schema theorem.
- Add symbol `\*' to binary alphabet. It matches both 0 and 1
- Assume fixed symbols are part of the solution, \* is don't care
- Order of a schema *o(S)*: number of fixed symbols
- Defining length of a schema  $\delta(S)$ : the distance between the first and the last fixed symbols
- Fitness proportionate selection. Crossover (with  $p_c$ ) and mutation (with  $p_m$ ) assumed to be destructive
- Number of matches for *S* at time *t* in the pool:  $\xi(S; t)$

• 
$$\xi(S, t+1) \ge \frac{\xi(S, t) f(S)}{f_{avg}} \left( 1 - \frac{p_c \delta(S)}{(l-1)} - p_m o(S) \right)$$

#### Schema Theorem:

Short, low-order, above-average schemata receive exponentially increasing trials in subsequent generations of a genetic algorithm.

#### Building Block Hypothesis:

A genetic algorithm seeks near-optimal performance through the juxtaposition of short, low-order, high performance schemata called the building blocks.

## **Crossover is destructive?**

- Schema theorem assumes mutation and crossover are destructive in a schema. If both parents do not have the same schemata this assumption is correct.
- Find a better schema as a result of genetic operator is possible.
- Mixing building blocks:

Combining two building blocks together in an individual as a result of crossover.

Crossover of (1\*\*\*1\*\*\*\*) and (\*\*\*\*\*0\*0\*) at position
 5 results in combination of two building blocks in one of the offsprings, (1\*\*\*1\*0\*0\*)

# **Mixing Building Blocks**

- Single point crossover analysis
- Assume real value length 1 chromosomes containing infinite number of genes. (each gene is a point in the continuous line of chromosomes)
- Cases are:
  - Case 1

Elements of the both building block are tightly placed.

- Case 2

Elements of the one building block are tightly placed the other is randomly placed.

Case 3
 Both elements are randomly placed

• Assume tight building blocks are single points. Mixing probabilities for crossover point  $\alpha$ 



$$EV_{tight-tight}(k) = \int_{0}^{1} 2\alpha(1-\alpha)d\alpha = \frac{1}{3}$$

$$EV_{tight-random}(k) = \int_{0}^{1} \alpha^{k} (1-\alpha) + \alpha (1-\alpha)^{k} d\alpha = \frac{2}{(k+1)(k+2)}$$

$$EV_{random-random}(k) = \int_{0}^{1} 2\left[\alpha^{k}(1-\alpha)^{k}\right] d\alpha = \frac{2(k!)^{2}}{(2k+1)!}$$



#### Schema Context

- Fixed positions in each schema defines a partitioning of the search space {\*0\*\*0\*, \*0\*\*1\*, \*1\*\*0\*, \*1\*\*1\*}
- When a building block dominates the population, the search space is reduced to one of these partitions.
- What about non-fixed positions? Their frequency, how BB's interact with the others? Context of a schema
- Context of schema H is defined as a set of conditional probability values  $p_i$  for each individual in schema H.

$$C(H) = \left\{ p'_{i} = \frac{p_{i}}{\sum_{j \in H} p_{j}}, \quad i \in H \right\}$$

• Flat context: all individuals have equal probability.  $C(H) = \left\{ p'_i = 2^{k-l}, i \in H \right\}$ 

#### Context in a population:

 $m_i$ : number of individual representing schema *i* in population *P* 

$$C_P(H) = \begin{cases} p'_i = \frac{m_i}{\sum_{j \in H} m_j}, & i \in H \end{cases}$$

• **Converged population:** schemata come from the same individual *I*.

$$C(H) = \{ p'_i = 1, i \in I ; p'_i = 0 \text{ otherwise} \}$$

## **BB** Superiority

- Survival of the BB depends on if it is superior in the population context.
- Assume: f(0\*\*\*\*) > f(1\*\*\*\*), f(00\*\*\*\*) > f(01\*\*\*\*) > f(10\*\*\*\*) > f(11\*\*\*\*) but f(111\*\*\*) is greater than all others.
- *f*(111\*\*\*) is locally superior BB of solution 111111
- Deceptive problems: problems having superiority of low order schema favors a sub-optimal solution.

#### **Deceptive Problems**

• Example, a six bit trap function



- All lower order schema cause 000000 to converge.
- Some problems are deceptive in nature. Like problems with high ephistasis (semantic interaction among genes).

# Linkage Learning

- Genes semantically linked however positioned not close in genotype have a smaller probability of survival and mixing.
- Linkage learning methods try to discover the linkage among the genes so that they treated specially during recombination.
- Messy GA (mGA) by Goldberg (1989) introduces a variable length position free encoding for adapting/preserving linkage. Other methods include:
  - LLGA (Harik)
  - Symbiotic Evolution
  - BOA (Pelikan), estimation of distribution.

## Messy GA

- Messy encoding. Keep the position of gene in the chromosome together with its value:
   (3,0) (2,0) (5,1) (4,0) (1,1) = 10001
- Problems: under-specification and over-specification
- Over-specification:
  - Majority voting, majority wins. Deception favored?
  - First come first served. Positional priority.
- Under-specification. Harder problem:
  - Random pick
  - k-bit perturbations. Start from k=1 and as long as k+1 achieves better fitness, use it for k+1.

- Crossover. Cut and splice operations on chromosomes.
- Phases of mGA
  - Initialization: generation of all sub-strings of a certain length k
  - Primordial: selection of these enumerated building blocks against a particular template.
  - Juxtapositional: selection, cut, splice, and usual GA operations.
- A building block order k is assumed at each epoch. Start with k=1 to find a competitive template. Increase k at each epoch.
- Fast mGA, a modification of mGA to make operations more efficient.

## Forces in Mixing

- Phenomena occurring in EC that affects the proper convergence:
  - Takeover: whole population consists of a single superior individual
  - Drift: random takeover any individual.
  - Cross-competition of BB's: BB's from different individuals not superior to each other end of with a poor representation of the solution.
- Mixing occurs if it finds a chance among these forces.

## Forces in Mixing: Takeover

- Takeover time analysis. Use s tournament selection. Try to find # of generations for a superior individual to have copies through the whole the population.
- *n* tournaments of size *s*. Each superior individual can be selected with probability *s/n*. We expect to see it *s* times larger frequency at each generation.

$$s^{t_s} = n$$
$$t_s = \frac{\ln(n)}{\ln(s)}$$

• Mixing should occur before this.  $t_s > c t_{mix}$ 

### Forces of Mixing: Mixing time

 Mixing probability of two BB's of size k with uniform crossover (ignore the linkage).

 $p_{\text{mix}} = 2\left(\frac{1}{2}\right)^{2k}$ , if some bits are common ,  $\mu \le 2$   $p_{\text{mix}} = \frac{2}{2^{\mu k}}$ 

 If (n/2)p<sub>c</sub> recombinations in one generation. Mixing time, the number of generations required to get a mixing is:

$$t_{\rm mix} = \frac{1}{(n/2) p_c p_{\rm mix}} = \frac{2^{\mu k}}{n p_c}$$

• Mixing time should be less than take over time:

$$t_{s} > c t_{mix} \quad \frac{\ln n}{\ln s} > c \frac{2^{\mu k}}{n p_{c}} \qquad n \ln(n) > c \frac{2^{\mu k} \ln(s)}{p_{c}}$$

## Forces of Mixing: Drift

- Even with no selection pressure. Traits of an individual can have a random increase in population that is called drift.
- Drift time is proportional to the population size.  $t_d = c'n$
- Drift time should be less than takeover time:

$$t_d < t_s \quad \ln(s) < \frac{\ln(n)}{c'n}$$

• Not a practical consideration for *n* is sufficiently large.

## Forces of Mixing: Cross-competition

 Selection gives s copies of each best individual. After first generation, n/s independent individuals which can construct better individuals.

 $p_{0}: \text{ ratio of bit positions already fixed correctly}$  $(1-p_{0})^{n/s}: \text{ correct alleleles that does not exist in the population}$  $[1-(1-p_{0})^{n/s}]l \text{ alleles covered by the population}$  $(1-\alpha)l < [1-(1-p_{0})^{n/s}]l , \alpha: \text{failure ratio}$  $s > \frac{n \ln(1-p_{0})}{\ln(\alpha)}$ 

 If selection pressure is a significant fraction of the population size, we may expect to have cross-competition failure.

### **Forces of Mixing Revisited**

- Takeover vs. mixing:  $n \ln(n) > c \frac{2^{\mu k} \ln(s)}{p_c}$
- Drift:

$$\ln(s) < \frac{\ln(n)}{c'n}$$

• Cross-competition:

$$s < \frac{n \ln(1-p_0)}{\ln(\alpha)}$$

#### **Mixing Results**

• Thierens, 1999, EC 7(4).



• Selection pressure vs. crossover probability.



• Building block size vs. population size



• Goldberg, Deb, Thierens, 1992.

