# **Crossover Techniques in GAs**

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# **Reproduction in Genetic Algorithm**

#### Reproduction:

- Crossover
- Mutation
- Inversion

These genetic operators varies from one encoding scheme to another.

- Binary coded GAs
- Real-coded GAs
- Tree-coded GAs

# **Mating Pool: Prior to crossover operation**

- A mating pair (each pair consists of two strings) are selected at random. Thus, if the size of mating pool is N, then <sup>N</sup>/<sub>2</sub> mating pairs are formed.[Random Mating]
- The pairs are checked, whether they will participate in reproduction or not by tossing a coin, whose probability being p<sub>c</sub>. If p<sub>c</sub> is head, then the parent will participate in reproduction. Otherwise, they will remain intact in the population.

#### Note:

Generally,  $p_c = 1.0$ , so that almost all the parents can participate in production.



# **Crossover operation**

Once, a pool of mating pair are selected, they undergo through crossover operations.

- In crossover, there is an exchange of properties between two parents and as a result of which two offspring solutions are produced.
- The crossover point(s) (also called k-point(s)) is(are) decided using a random number generator generating integer(s) in between 1 and L, where L is the length of the chromosome.
- Then we perform exchange of gene values with respect to the k-point(s)

There are many exchange mechanisms and hence crossover strategies.



#### **Crossover Techniques in Binary Coded GA**

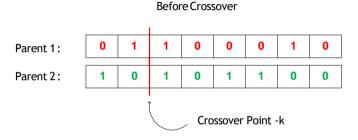
#### **Crossover operations in Binary-coded GAs**

- There exists a large number of crossover schemes, few important of them are listed in the following.
  - Single point crossover
  - Two-point crossover
  - Multi-point crossover (also called n-point crossover)
  - Uniform crossover (UX)
  - Half-uniform crossover (HUX)
  - Shuffle crossover
  - Matrix crossover (Tow-dimensional crossover)
  - Three parent crossover

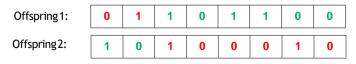
# Single point crossover

- $\blacksquare$  Here, we select the K-point lying between 1 and L. Let it be k.
- A single crossover point at k on both parent's strings is selected.
- All data beyond that point in either string is swapped between the two parents.
- The resulting strings are the chromosomes of the offsprings produced.

# Single point crossover: Illustration



#### Select crossover points randomly

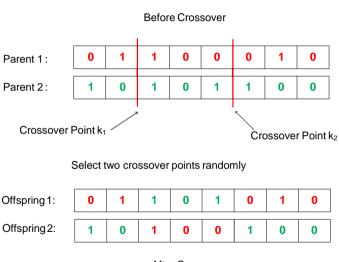


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#### **Two-point crossover**

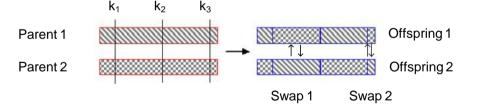
- In this scheme, we select two different crossover points  $k_1$  and  $k_2$  lying between 1 and L at random such that  $k_1 f = k_2$ .
- The middle parts are swapped between the two strings.
- Alternatively, left and right parts also can be swapped.

#### **Two-point crossover: Illustration**



#### **Multi-point crossover**

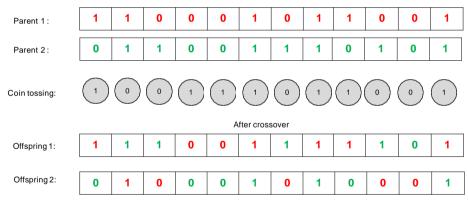
- In case of multi-point crossover, a number of crossover points are selected along the length of the string, at random.
- The bits lying between alternate pairs of sites are then swapped.



# **Uniform Crossover (UX)**

- Uniform crossover is a more general version of the multi-point crossover.
- In this scheme, at each bit position of the parent string, we toss a coin (with a certain probability  $p_s$ ) to determine whether there will be swap of the bits or not.
- The two bits are then swapped or remain unaltered, accordingly.

# **Uniform crossover (UX): Illustration**



Before crossover

Rule: If the toss is 0 than swap the bits between P1 and P2

#### Uniform crossover with crossover mask

- Here, each gene is created in the offspring by copying the corresponding gene from one or the other parent chosen according to a random generated binary crossover mask of the same length as the chromosome.
- Where there is a 1 in the mask, the gene is copied from the first parent
- Where there is a 0 in the mask, the gene is copied from the second parent.
- The reverse is followed to create another offsprings.

# Uniform crossover with crossover mask: Illustration



Offspring 1:

Offspring 2:

0

0 After Crossver

0

0

0

0

1

0

When there is a 1 in the mask, the gene is copied from Parent 1 else from Parent 2.

When there is a 1 in the mask, the gene is copied from Parent 2 else from Parent 1.

#### Half-uniform crossover (HUX)

- In the half uniform crossover scheme, exactly half of the non-matching bits are swapped.
  - Calculate the Hamming distance (the number of differing bits) between the given parents.
  - 2 This number is then divided by two.
  - The resulting number is how many of the bits that do not match between the two parents will be swapped but probabilistically.
  - Choose the locations of these half numbers (with some strategies, say coin tossing) and swap them.

#### Half-uniform crossover: Illustration

#### Before crossover

Parent 1:

1	1	0	0	0	0	1	0
1	0	0	1	1	0	1	1

Here, Hamming distance is 4

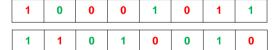
Tossing:



If toss is 1, then swap the bits else remain as it is

Offspring 1:

Offspring 2:

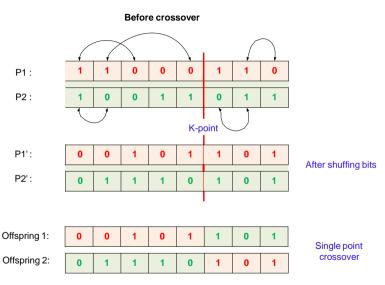


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#### Shuffle crossover

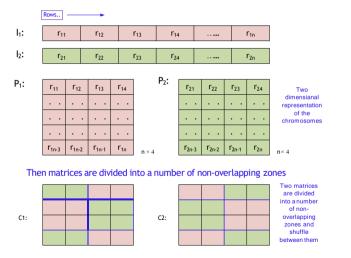
- A single crossover point is selected. It divides a chromosome into two parts called schema.
- In both parents, genes are shuffled in each schema. Follow some strategy for shuflling bits
- Schemas are exchanged to create offspring (as in single crossover)

#### **Shuffle crossover: Illustration**



#### **Matrix crossover**

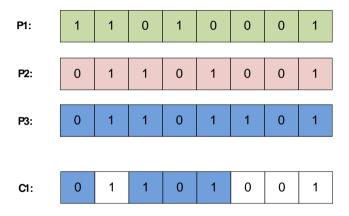
The matrix crossover strategy is expained with the following illustration.



#### Three parent crossover

- In this techniques, three parents are randomly chosen.
- Each bit of the first parent is compared with the bit of the second parent.
- If both are the same, the bit is taken for the offspring.
- Otherwise, the bit from the third parent is taken for the offspring.

#### Three parent crossover: Illustration



**Note:** Sometime, the third parent can be taken as the crossover mask.



# Comments on the binary crossover techniques

#### Non-uniform variation:

It can not combine all possible schemas (i.e. building blocks)

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For example: it can not in general combine instances of 11 ** ** * 1 and ** ** * 11 * * to form an instance of 11 ** 11 * 1.
```

#### Positional bias:

The schemas that can be created or destroyed by a crossover depends strongly on the location of the bits in the chromosomes.

# **Comments on the binary crossover techniques**

#### End-point bias:

It is also observed that single-point crossover treats some loci preferentially, that is, the segments exchanged between the two parents always contain the end points of the strings.

#### Hamming cliff problem:

A one-bit change can make a large (or a small) jump.

A multi-bits can make a small (or a large gap).

For example,  $1000 = \Rightarrow 0111$ 

(Here, Hamming distance = 4, but distance between phenotype is 1)

Similarly,  $0000 = \Rightarrow 1000$ 

(Here, Hamming distance = 1, but distance between phenotype is 8)



# Comments on the binary crossover techniques

- To reduce the positional bias and end-point bias, two-point crossover and multi-point crossover schemes have been evolved.
- In contrast, UX and HUX distribute the patterns in parent chromosomes largely resulting too much deflections in the offspring.
- In summary, binary coding is the simplest encoding and its crossover techniques are fastest compared to the crossover techniques in other GA encoding schemes.

Thank you.