



# Chapter 3

## GAs: Why Do They Work?

# Schema Theorem

- **SGA's features:**
  - binary encoding
  - fitness-proportional selection
  - one-point crossover
  - bit-flipping with fixed probability mutation
- Schema Theory assures that the search process carried out by a genetic algorithm develops in the direction of the proliferation of the most successful schemas/schemata, concomitantly with destroying unsuccessful schemas.

# Definitions

- A **schema** represents all strings, which match it on all positions other than “\*”. The number of “\*” in a schema determines the number of strings matched by the schema.
  - There are  $3^m$  schemas for a  $m$ -length string.
  - A schema matches  $2^r$  strings. ( $r$ : # of “\*”)
  - A string is matched by  $2^m$  schemas. ( $m$ : the length of the string)

- $o(S)$ , the **order** of a schema  $S$ , is the number of 0 and 1 positions.
- $\delta(S)$ , the **defining length** of the schema  $S$ , is the distance between the first and the last fixed string positions.
- $\xi(S, t)$  is the **number of strings** in a population at the time  $t$ , matched by schema  $S$ .
- $eval(S, t)$  is the **average fitness** of all strings in the population at the time  $t$ , matched by the schema  $S$ .
- $\overline{F(t)}$ , **the average performance** (or average fitness) of the entire population at the time  $t$ .

- $\xi(S, t)$  is the **number of strings** in a population at the time  $t$ , matched by schema  $S$ .

$S = (****111*****)$

```

v1 = (10011010000000111111010011011111)
v2 = (111000100100110111001010100011010)
v3 = (000010000011001000001010111011101)
v4 = (100011000101101001111000001110010)
v5 = (000111011001010011010111111000101)
v6 = (000101000010010101001010111111011)
v7 = (001000100000110101111011011111011)
v8 = (100001100001110100010110101100111)
v9 = (010000000101100010110000001111100)
v10 = (000001111000110000011010000111011)
v11 = (011001111110110101100001101111000)
v12 = (110100010111101101000101010000000)
v13 = (111011111010001000110000001000110)
v14 = (010010011000001010100111100101001)
v15 = (111011101101110000100011111011110)
v16 = (110011110000011111100001101001011)
v17 = (011010111111001111010001101111101)
v18 = (011101000000001110100111110101101)
v19 = (000101010011111111110000110001100)
v20 = (101110010110011110011000101111110)

```

- $eval(S, t)$  is the **average fitness** of all strings in the population at the time  $t$ , matched by the schema  $S$ .

$$eval(S, t) = \sum_{j=1}^{\rho} eval(v_{t_j}) / \rho$$

$$\text{where } \rho = \xi(S, t)$$

- $\overline{F(t)}$  , **the average performance** (or average fitness) of the entire population at the time  $t$ .

$$\overline{F(t)} = \frac{F(t)}{\text{populatio\_size}}$$

where

$$F(t) = \sum_{j=1}^{\text{pop\_size}} \text{eval}(v_{t_j})$$

# Effects on Schema Dynamics

1. Effect of selection

2. Effect of crossover

- Effect of selection & crossover

3. Effect of mutation

- Effect of selection & crossover & mutation



$$S = (****111*****))$$

*t*

$v_1 = (10011010000000111111010011011111)$   
 $v_2 = (111000100100110111001010100011010)$   
 $v_3 = (000010000011001000001010111011101)$   
 $v_4 = (100011000101101001111000001110010)$   
 $v_5 = (000111011001010011010111111000101)$   
 $v_6 = (000101000010010101001010111111011)$   
 $v_7 = (001000100000110101111011011111011)$   
 $v_8 = (100001100001110100010110101100111)$   
 $v_9 = (010000000101100010110000001111100)$   
 $v_{10} = (000001111000110000011010000111011)$   
 $v_{11} = (011001111110110101100001101111000)$   
 $v_{12} = (110100010111101101000101010000000)$   
 $v_{13} = (111011111010001000110000001000110)$   
 $v_{14} = (010010011000001010100111100101001)$   
 $v_{15} = (111011101101110000100011111011110)$   
 $v_{16} = (110011110000011111100001101001011)$   
 $v_{17} = (011010111111001111010001101111101)$   
 $v_{18} = (011101000000001110100111110101101)$   
 $v_{19} = (000101010011111111110000110001100)$   
 $v_{20} = (101110010110011110011000101111110)$

*t+1*

$v'_1 = (011001111110110101100001101111000) (v_{11})$   
 $v'_2 = (100011000101101001111000001110010) (v_4)$   
 $v'_3 = (001000100000110101111011011111011) (v_7)$   
 $v'_4 = (011001111110110101100001101111000) (v_{11})$   
 $v'_5 = (0001010100111111111110000110001100) (v_{19})$   
 $v'_6 = (100011000101101001111000001110010) (v_4)$   
 $v'_7 = (111011101101110000100011111011110) (v_{15})$   
 $v'_8 = (000111011001010011010111111000101) (v_5)$   
 $v'_9 = (011001111110110101100001101111000) (v_{11})$   
 $v'_{10} = (000010000011001000001010111011101) (v_3)$   
 $v'_{11} = (111011101101110000100011111011110) (v_{15})$   
 $v'_{12} = (010000000101100010110000001111100) (v_9)$   
 $v'_{13} = (000101000010010101001010111111011) (v_6)$   
 $v'_{14} = (100001100001110100010110101100111) (v_8)$   
 $v'_{15} = (101110010110011110011000101111110) (v_{20})$   
 $v'_{16} = (111001100110000101000100010100001) (v_1)$   
 $v'_{17} = (111001100110000100000101010111011) (v_{10})$   
 $v'_{18} = (111011111010001000110000001000110) (v_{13})$   
 $v'_{19} = (111011101101110000100011111011110) (v_{15})$   
 $v'_{20} = (110011110000011111100001101001011) (v_{16})$

# Effects on Schema Dynamics

## 1. Effect of selection

$$\begin{aligned}\zeta(S, t + 1) &= \zeta(S, t) \cdot \text{eval}(s, t) / \overline{F(t)} \\ &= \zeta(S, t) \cdot (1 + \varepsilon)\end{aligned}$$

### Conclusion:

1.  $\varepsilon > 0$  **above average** schema; exponentially increasing  
(An above average schema, S, receives an exponentially increasing number of strings in the next generations.)
2.  $\varepsilon = 0$  **average schema**; staying the same level  
(An average schema, S, stays on the same level.)
3.  $\varepsilon < 0$  **below average** schema; exponentially decreasing  
(A below average schema, S, receives an decreasing number of strings in the next generations.)

# Effects on Schema Dynamics

## 2. Effect of selection & crossover

$$\begin{aligned}\zeta(S, t+1) &\geq \left\lfloor \zeta(S, t) \cdot \text{eval}(S, t) / \overline{F(t)} \right\rfloor \left[ 1 - p_c \frac{\delta(S)}{m-1} \right] \\ &= \zeta(S, t) \cdot (1 + \varepsilon)\end{aligned}$$

### Conclusion

The **short , above-average** schema will receive an exponentially increasing number of strings in the next generation.

# Effects on Schema Dynamics

## 3. Effect of selection & crossover & mutation

$$\begin{aligned}\zeta(S, t+1) &\geq \left[ \zeta(S, t) \cdot \text{eval}(S, t) / \overline{F(t)} \right] \left[ 1 - p_c \frac{\delta(S)}{m-1} \right] [1 - o(S)p_m] \\ &\cong \left[ \zeta(S, t) \cdot \text{eval}(S, t) / \overline{F(t)} \right] \left[ 1 - \rho_c \frac{\delta(S)}{m-1} - o(S)\rho_m \right] \\ &= \zeta(S, t) \cdot (1 + \varepsilon)\end{aligned}$$

### Conclusion :

The **above-average** schemata with **short** defining length and **low-order** will be sampled at exponentially increased rate.

# Conclusion

## **Theorem 1** (Schema Theorem):

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

## **Hypothesis 1** (Building Block Hypothesis):

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