

Lab 1 - Implementing A Simple Evolutionary Algorithm

CSE, SUSTech

Outline of This Lecture

- Teaching Assistant
- A Simple Evolutionary Algorithm (EA)
- How Does a Simple EA Work
- Highlights of This Simple EA
- Does The Population Size Matter?
- Illustrate The Results!

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A Simple Evolutionary Algorithm (EA)

1. Generate the initial population $P(0)$ at random;
2. Set $i \leftarrow 0$; // Generation counter
3. REPEAT
 - a. Evaluate the fitness of each individual in $P(i)$;
 - b. Select parents from $P(i)$ based on their fitness in $P(i)$;
 - c. Generate offspring from the parents using crossover and mutation to form $P(i + 1)$;
 - d. $i \leftarrow i + 1$;
4. UNTIL halting criteria are satisfied

How Does a Simple EA Work - Example

Let's use the simple EA to maximise the function

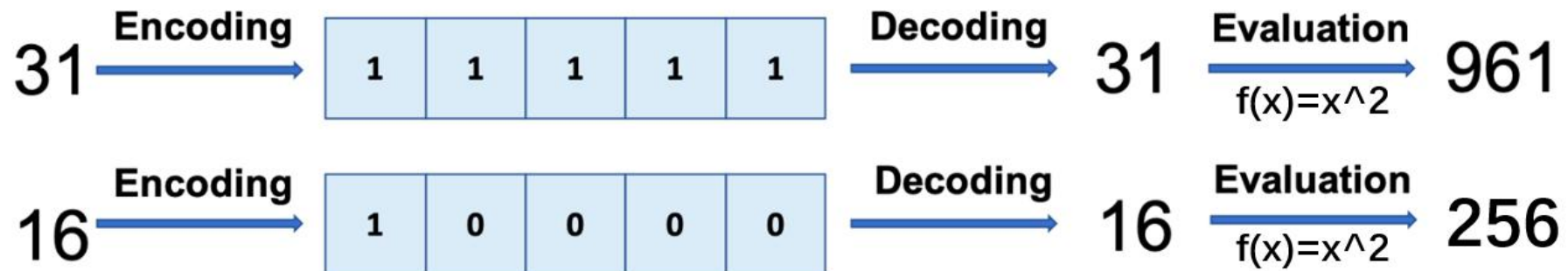
$$f(x) = x^2$$

with x in the integer interval $[0, 31]$, i.e., $x = 0, 1, \dots, 30, 31$.

How Does a Simple EA Work - Representation

The first step of EA applications is encoding, i.e., the representation of chromosomes:

- We adopt binary representation for integers.
- Five bits are used to represent integers up to 31.



How Does a Simple EA Work I

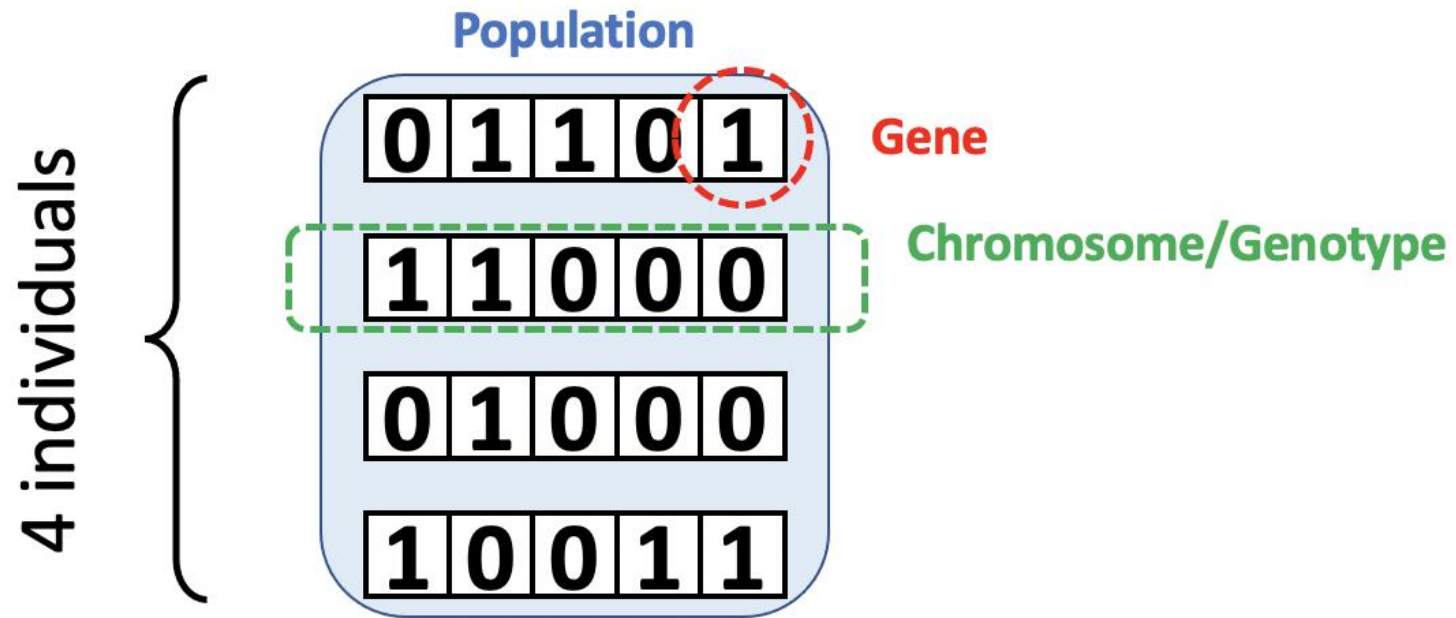
Assume that the population size is 4.

1. Generate initial population at random, e.g.,

01101, 11000, 01000, 10011.

These are chromosomes (染色体) or genotypes (基因型).

How Does a Simple EA Work II



How Does a Simple EA Work III

2. Calculate fitness value for each individual.

a) Decode the individual into an integer (called phenotypes 表现型),

01101 \rightarrow 13

11000 \rightarrow 24

01000 \rightarrow 8

10011 \rightarrow 19

b) Evaluate the fitness according to $f(x) = x^2$,

01101 \rightarrow 13 \rightarrow 169

11000 \rightarrow 24 \rightarrow 576

01000 \rightarrow 8 \rightarrow 64

10011 \rightarrow 19 \rightarrow 361

How Does a Simple EA Work IV

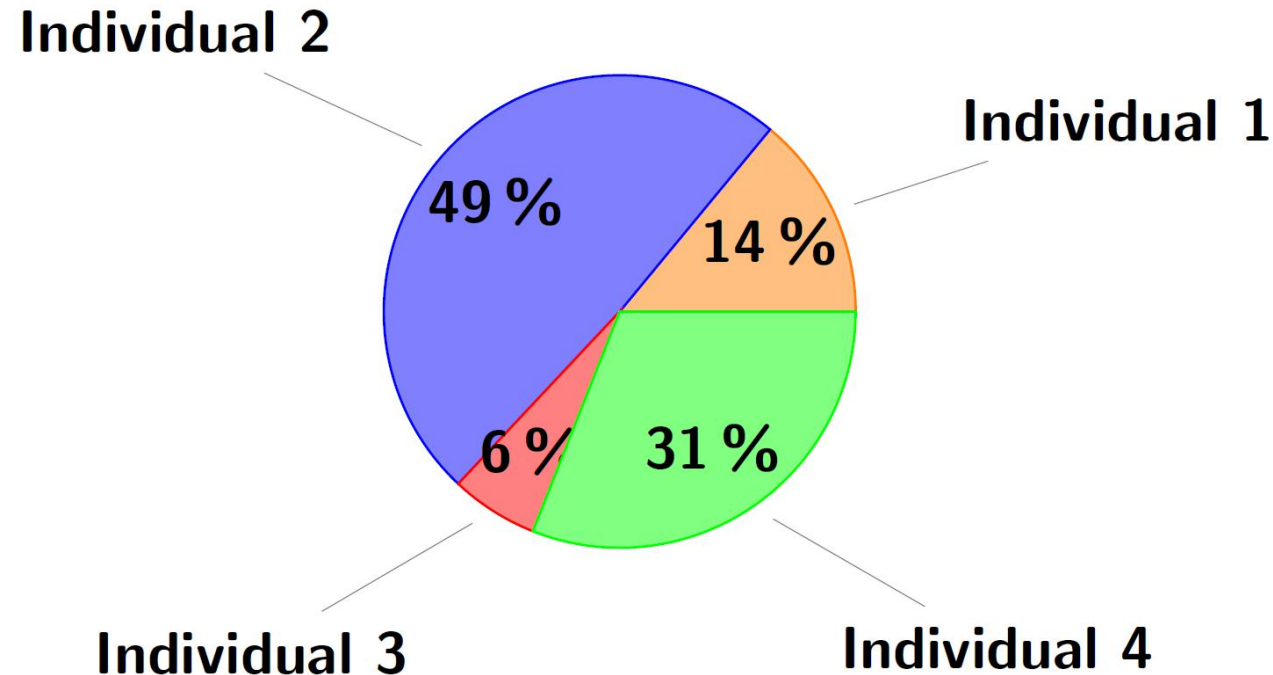
3. Select two individuals for crossover based on their fitness.
 - a) If roulette-wheel selection is used, then

$$p_i = \frac{f_i}{\sum_j f_j}.$$

Two offspring are often produced and added to an intermediate population. Repeat this step until the intermediate population is filled. In our example,

$$\begin{aligned} p_1(13) &= 169/1170 = 0.14 & p_2(24) &= 576/1170 = 0.49 \\ p_3(8) &= 64/1170 = 0.06 & p_4(19) &= 361/1170 = 0.31 \end{aligned}$$

How Does a Simple EA Work V



- b) Assume we have `crossover(01101, 11000)` and `crossover(10011, 11000)`. We may obtain offspring 01100 and 11001 from `crossover(01101, 11000)` by choosing a random crossover point at 4, and obtain 10000 and 11011 from `crossover(10011, 11000)` by choosing a random crossover point at 2. Now the intermediate population is 01100, 11001, 10000, 11011.

How Does a Simple EA Work VI

4. Apply mutation to individuals in the intermediate population with a small probability. A simple mutation is bit-flipping. For example, we may have the following new population $P(1)$ after random mutation:

01101, 11001, 00000, 11011

The initial population is:

01101, 11000, 01000, 10011.

5. Goto Step 2 if not stop.

Highlights of This Simple EA

- Binary representation of individuals.
- Roulette-wheel selection.
- One point crossover.
- Bit-flipping mutation.
- Replacement: the current population will be replaced by the new population of the same size.

Does The Population Size Matter?

- Let's try the following: vary the population size:
 - ✓ Vary the population size $\mu = 4, 10$.
 - ✓ At each generation, record the best solution in the current population and its fitness.
 - ✓ At each generation, record the best-so-far solution and its fitness.

Illustrate The Results!

- Plot 4 figures:
 - ✓ x-axis: current evaluation number. y-axis: best-so-far fitness value.
 - ✓ x-axis: current evaluation number. y-axis: fitness value of the best individual of current population.