

Genetic Algorithm

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Outline

- Background: Biological Genetics & GA
- Two Examples
- Some Applications
- Online Demos* (if the time allows)

Introduction

● What is Genetic Algorithms (GA)?

- A **search technique** used in computing to find true or approximate solutions to optimization and search problems (Wikipedia).
- Suitable for searching problems with extremely huge search space that are hard or even impossible to be solved by traditional analytical methods or minimum-seeking algorithms (i.e. exhaustive search or calculus based optimization method)
- Inspired by Darwin's theory of evolution and the study of genetics such as inheritance, mutation, selection, and crossover (also called recombination), etc.
- Originally invented by John Holland from the University of Michigan in the early seventies(1975)

Biological Background

● Organism:

- Cell – the basic part of all living organism
- Chromosome – each cell contains the same number of chromosomes
- Gene – a chromosome consists of genes, the traits are encoded in the form of DNAs or genetic codes

● Reproduction:

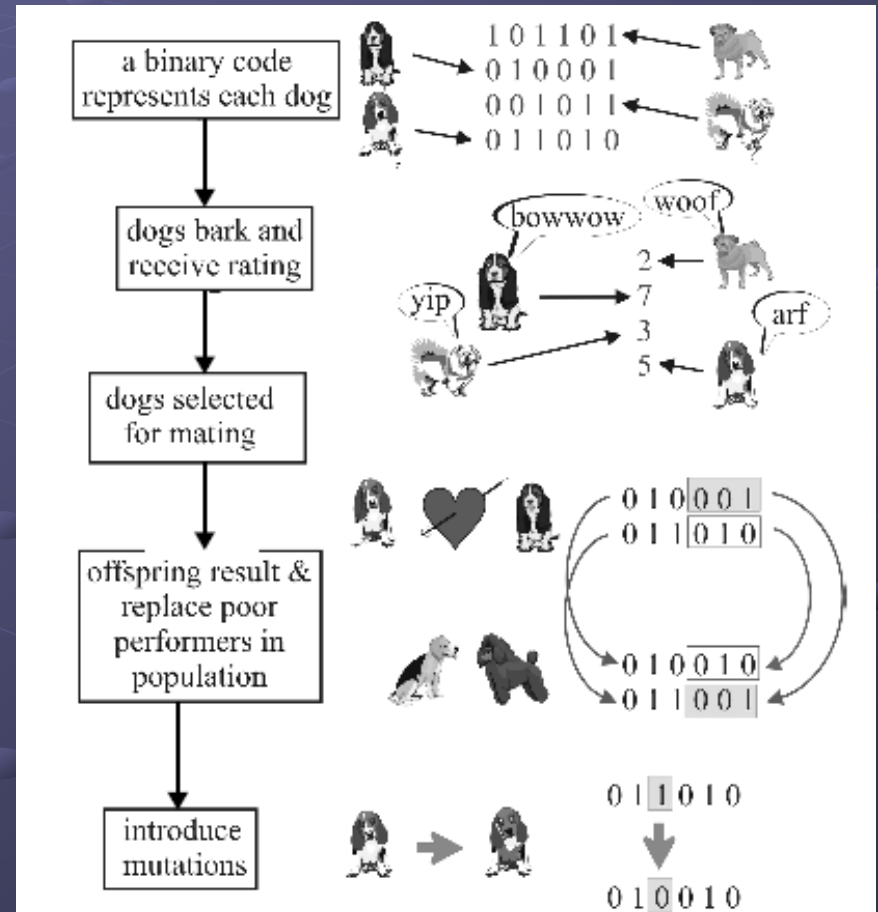
- Inheritance – offspring inherits the traits (genetic code) from parents
- Crossover – the portion of mother's and father's chromosomes
- Mutation – a random change may occurs in the offspring's genetic code

● Natural selection & evolution (Darwin's theory)

- Fitness – how successful is an individual fit for the environment, determined by genetic code.
- Natural selection- higher fitness value means higher survive rate
- Evolution – on the long run, the most recent survived generation intend to have a high fitness

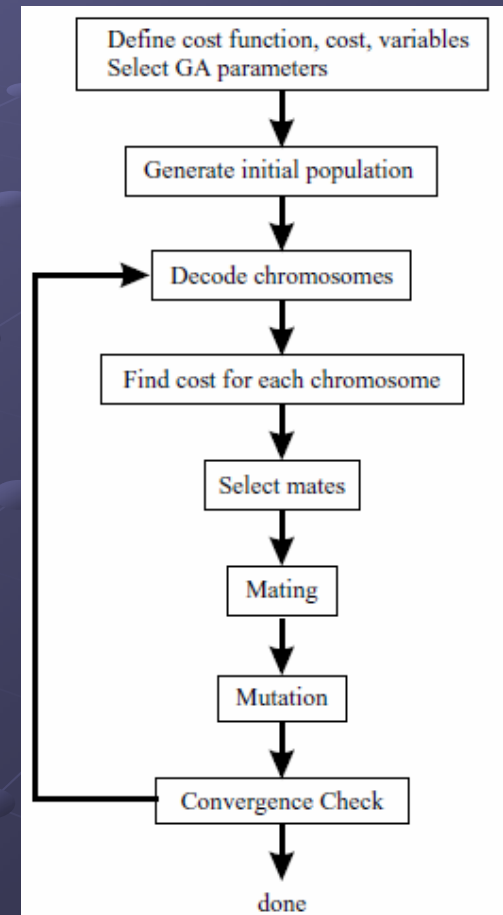
GA v.s Biological Genetics

- GA is the simulation of the natural selection and evolution on a computer
- An intuitive analogy



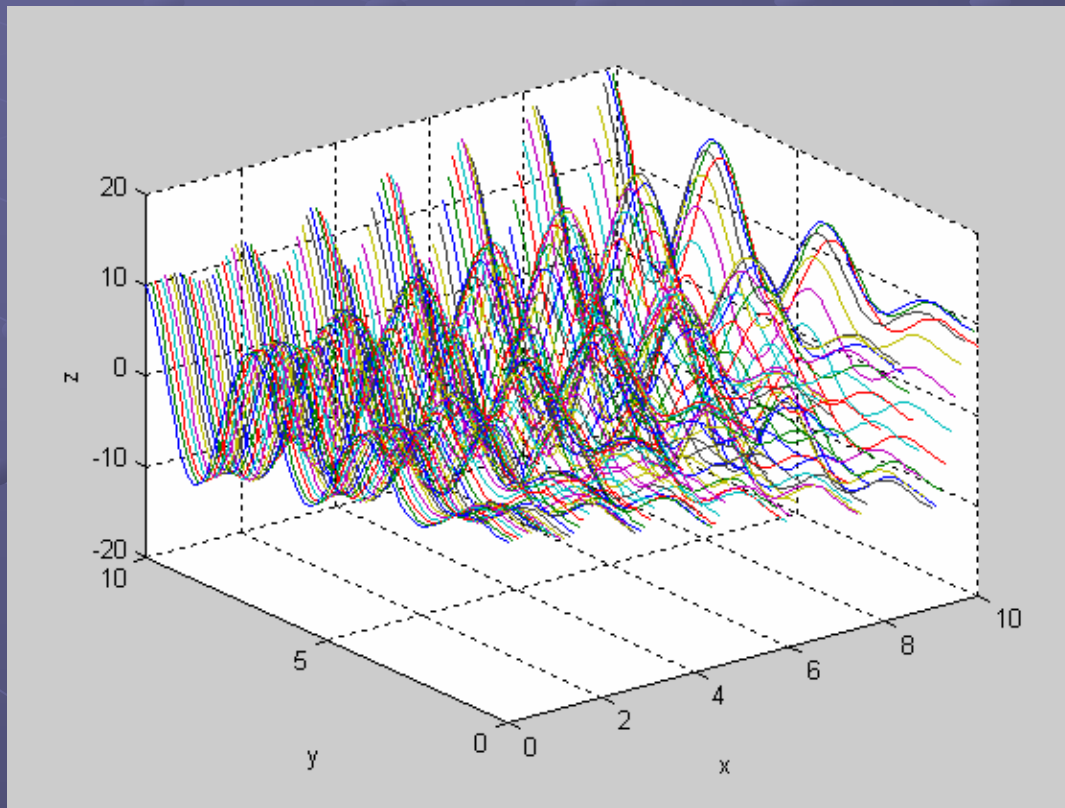
Flowchart

- The GA starts with an initial population that consists of a set of individuals, each individual is represented by a unique binary string or float-point value (chromosome).
- Individuals are evaluated (based on their fitness), selected, and mated
- New individuals are generated in the principles of inheritance and variation. Individuals with low fitness are discarded.
- The iteration continues until a convergence requirement is fulfilled



A Basic Example

- Problem: find the global maximum point of the following surface (2D function):



$$f(x, y) = x\sin(4x) + 1.1y\sin(2y)$$
$$0 \leq x \leq 10, 0 \leq y \leq 10$$

GA Steps

- Chromosome Encoding: chromosome = [x, y]
- Fitness Function: $fitness = x \cdot \sin(4 \cdot x) + 1.1 \cdot y \cdot \sin(2 \cdot y)$
- First Generation: 16 random individuals (chromosomes)
- **Selection**: 8 most-fit chromosomes survive, others are discarded
- **Pairing**: 4 pairs of parents are randomly selected.
- **Mating & Crossover**: offspring is produced by parents:
 - blending method: $p_child = b \cdot p_pa + (1-b) \cdot p_ma$,
 - p is one of the variables (x or y), b is a random number between [0, 1], same or different b for each variable.
- **Mutation**: a few of randomly selected variables are replaced by uniform random numbers.
- New Generation: made up by survived parents and offspring
- Iteration: the above process is iterated until the global maximum is found, or the predefined iteration number is exceeded.

Parameters

● Elitism:

- One or a few best chromosomes are copied directly to the next generation, without changes

● Crossover probability: between [0, 1]

- 0%: all offspring is exact copy of parents (but the new generation is not necessarily the same!)
- 100%: all offspring is made by crossover

● Mutation probability: between [0, 1]

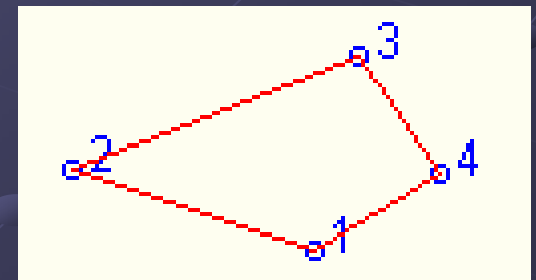
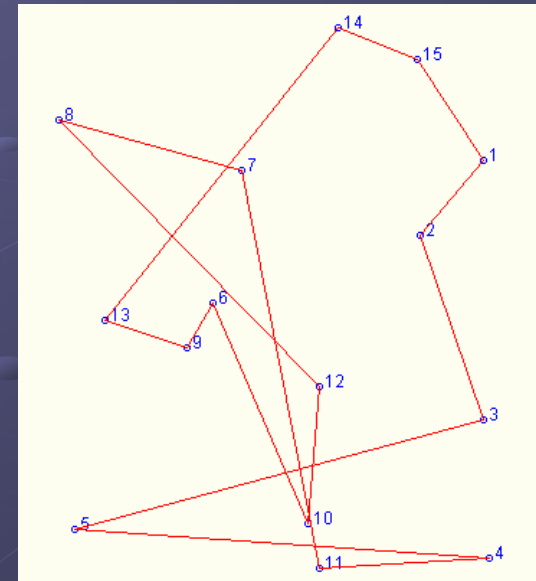
- 0%: no individual is changed
- 100%: all individuals are changed (random searching)

● Population Size:

- Too small: the search space is limited
- Too large: slows down

A Classical Example

- Traveling Salesman Problem (TSP):
 - Randomly located cities, the traveling salesman has to minimize the traveling distance while visit all of the cities
 - Essentially an ordering problem: finding an optimal ordering for a sequence of N items
 - Other similar problem: scheduling, routing, resource allocation, assignment
- A “natural” encoding method might be problematic!
 - The simplest case: four cities: 1, 2, 3, 4 – each with a unique bit-string encoding
 - Considering two individuals: 3-2-1-4, and 4-1-2-3
 - A crossover between the parents may produce wrong offspring, i.e. 3-2-2-3!
- “Random-key” encoding method [Forrest1993].
- In fact, for each specific GA problem, designing the encoding method is an important issue, so do the
 - crossover and mutation methods.
 - selection and pairing methods.
 - Refer to: <http://cs.felk.cvut.cz/~xobitko/ga/cromu.html>



Random-key Method*

- The chromosome is encoded by a bit string, which is divided into N segments of k bits [Forrest1993]
 - N is the same as the cities number
 - $2^k \gg N$
- For example: $N = 4$, $k = 4$ (16 bits for each chromosome)
- A randomly generated chromosome: [0101 0011 0001 1001], interpreted as decimal numbers: [5-3-1-9]
- City numbers are encoding as the segment position, sequence of traveling route is up to the relative value of each segment.
 - Segment 3 has the smallest value 1, so city 3 is the first.
 - Segment 2 has the second smallest value 3, so city 2 is the second, and so on and so forth
 - Thus, pattern [5-3-1-9] represents the tour [3-2-1-4]
- The offspring's legality is guaranteed, no matter what "strange" chromosome (bit string) is produced via crossover and/or mutation
- Also called permutation encoding

Online Demos*

- Global Maximum of 1D Function:

http://cs.felk.cvut.cz/~xobitko/ga/example_f.html

- Global Maximum of 2D Function:

<http://cs.felk.cvut.cz/~xobitko/ga/example3d.html>

- TSP:

<http://cs.felk.cvut.cz/~xobitko/ga/tspexample.html>

Examples in Music Tech

- Granular Synthesis Regulation [Fujinaga1994]
 - Each grain's parameters (freq, amp, etc.) mapped to a chromosome's bit string
 - GA parameters (crossover rate, mutation rate, etc.) control the change in the grain population
- FM Synthesis Parameters Optimization [Lai2006]
 - FM parameters (carrier freq, mod freq, amp) mapped to a chromosome's bit string
 - Spectrum similarity of target sound and synthesized sound is evaluated as fitness
- Automatic Generation of Sound Synthesis [Garcia2000]
 - Not only the parameters of synthesis, but the synthesis building block are regarded as genes
 - Generates new topologies and synthesizes the target sound
 - Genetic Programming
 - Tree coding
 - Lisp Implementation [Forrest1993]

Applications List

- General
 - Optimization (strategy planning, robot trajectory, sequence scheduling)
 - Modeling dynamic systems (ecologic systems, immune systems, genetic systems, social systems)
 - Evolving LISP programs (genetic programming)
 - Etc.
- Music Engineering / Composing
 - Synthesis system design
 - Synthesis parameters optimization
 - Sound synchronization, animation
 - Timbre recognition
 - Fugue generation
 - Jazz solos generation
 - Waveform evolving
 - Etc.
- Many others up to your imagination

References

- [Haupt2004] Haupt, Randy L., and Sue Ellen Haupt. 2004. *Practical Genetic Algorithms*. 2nd Edition ed. Hoboken, New Jersey: Wiley-Interscience.
- [Forrest1993] Forrest, S. 1993. Genetic algorithms: principles of natural selection applied to computation. *Science* 261 (5123):872-878.
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- [Fujinaga1994] Fujinaga, Ichino, and J Vantomme. 1994. Genetic Algorithms as a Method for Granular Synthesis Regulation. Paper read at International Computer Music Conference, 1994.
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Questions?

● Please ask...