

Εισαγωγή στους
Εξελικτικούς Αλγόριθμους

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Introduction to Evolutionary Computation

The EvoNet Flying Circus

Brought to you by *P. Adamidis*

The EvoNet Training Committee

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Q What is the most powerful
problem solver in the Universe?

A The (human) brain

that created "the wheel, New York, wars and so
on" (after Douglas Adams)

A The evolution mechanism

that created human brain (after Darwin et al.)

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**Building problem solvers by
looking at and mimicking:**

▪ brains → neurocomputing

▪ evolution → evolutionary computing

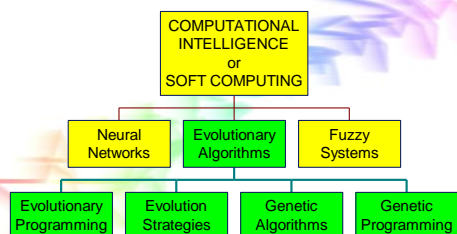
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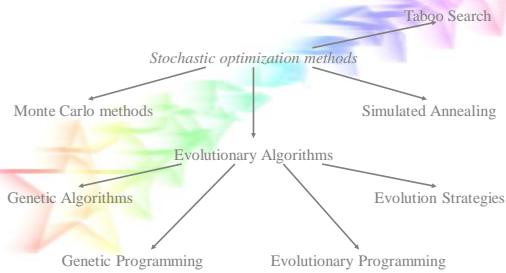
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Taxonomy



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Taxonomy (2)



History (1)



History (2)

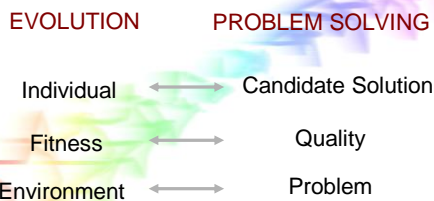
- L. Fogel 1962 (San Diego, CA): *Evolutionary Programming*
- J. Holland 1962 (Ann Arbor, MI): *Genetic Algorithms*
- I. Rechenberg & H.-P. Schwefel 1965 (Berlin, Germany): *Evolution Strategies*
- J. Koza 1989 (Palo Alto, CA): *Genetic Programming*

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History (3)

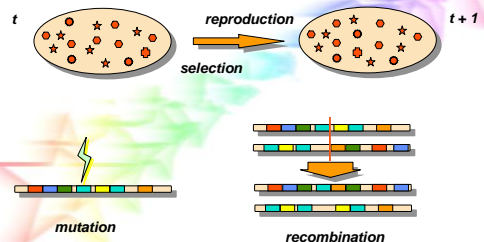
- 1859 Charles Darwin: inheritance, variation, natural selection
- 1957 G. E. P. Box: random mutation & selection for optimization
- 1958 Fraser, Bremermann: computer simulation of evolution
- 1964 Rechenberg, Schwefel: mutation & selection
- 1966 Fogel et al.: evolving automata - "evolutionary programming"
- 1975 Holland: crossover, mutation & selection - "reproductive plan"
- 1975 De Jong: parameter optimization - "genetic algorithm"
- 1989 Goldberg: first textbook
- 1991 Davis: first handbook
- 1993 Koza: evolving LISP programs - "genetic programming"

The Metaphor



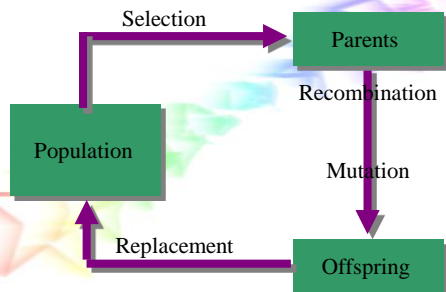
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The Ingredients



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The Evolutionary Cycle



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Pseudocode

```

generation = 0;
SeedPopulation(popSize); // at random or from a file
while(!TerminationCondition())
{
    generation = generation + 1;
    CalculateFitness(); // ... of new genotypes
    Selection(); // select genotypes that will reproduce
    Crossover( $p_{cross}$ ); // mate  $p_{cross}$  of them on average
    Mutation( $p_{mut}$ ); // mutate all the offspring with Bernoulli
                        // probability  $p_{mut}$  over genes
}
    
```

Evolution Programs

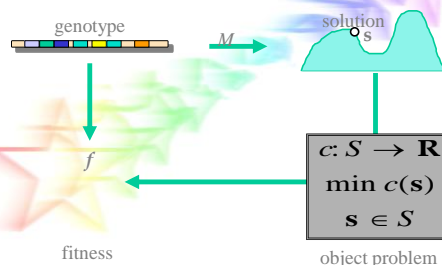
Slogan:

Genetic Algorithms + Data Structures = Evolution Programs

Key ideas:

- use a data structure as close as possible to object problem
- write appropriate genetic operators
- ensure that all genotypes correspond to feasible solutions
- ensure that genetic operators preserve feasibility

Object problem and Fitness



Κριτήρια Τερματισμού

- Μετά από ορισμένο αριθμό Γενεών
- Η ποιότητα του καλύτερου ατόμου δεν βελτιώνεται για ένα προκαθορισμένο αριθμό γενεών
- Εύρεση του ολικού βέλτιστου

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The Evolution Mechanism

- Increasing diversity by genetic operators
 - mutation
 - recombination
- Decreasing diversity by selection
 - of parents
 - of survivors

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Genetic-based Optimisation

- Η διαδικασία βελτιστοποίησης στηρίζεται:
 - προνομιακή επιβίωση και αναπαραγωγή του καλύτερου (preferential survival and reproduction of the fittest)
 - διατήρηση ενός πληθυσμού που παρουσιάζει ποικιλία (maintenance of a population with diverse members)
 - κληρονομικότητα της γενετικής πληροφορίας των γονέων (inheritance of genetic information from parents)
 - περιστασιακή μετάλλαξη γονιδίων (occasional mutation of genes)



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Domains of Application

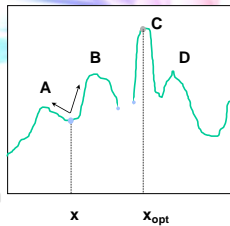
- Numerical, Combinatorial Optimisation
- System Modeling and Identification
- Planning and Control
- Engineering Design
- Data Mining
- Machine Learning
- Artificial Life

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Domains of Application (2)

□ Numerical optimization

- Difficult functions
 - Multimodal
 - Discontinuous
 - Nonlinear
- Parameter optimization
 - Structural optimization
 - Parametric design

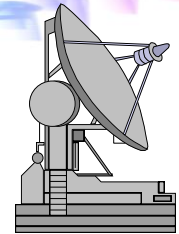


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Domains of Application (3)

■ Image processing and pattern recognition

- Satellite image processing
- Medical X-ray image processing
- Facial composites of criminal suspects

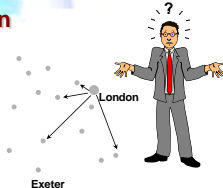


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Domains of Application (4)

□ Combinatorial optimization

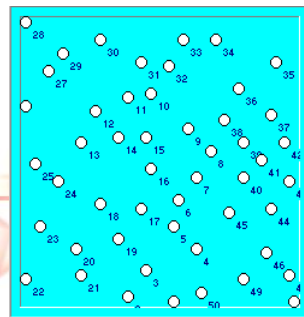
- Travelling salesperson
- Scheduling
- Networks
- Bin packing and pallet loading



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Travelling Salesperson

E x a m p l e

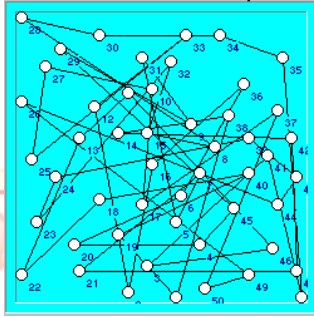


- 50 cities
- Size = 50!

TSP Solver Demo v. 1.0 by V.V. Myagkih, O.V. Savelev and V.M. Kureichik

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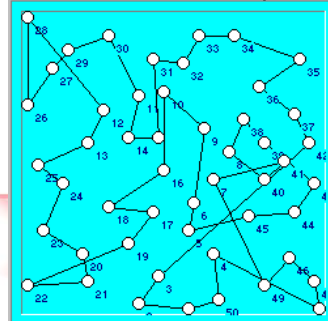
Travelling Salesperson (2)



- Random start
- Not very good

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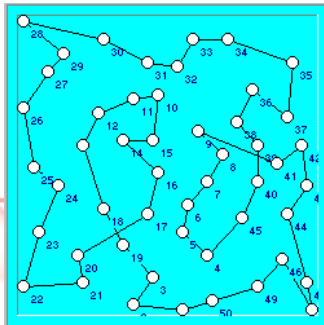
Travelling Salesperson (3)



- After 50 generations
- Better solution
- Not perfect

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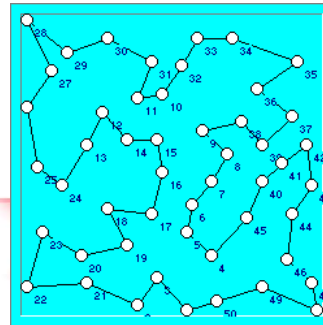
Travelling Salesperson (4)



- After 800 generations
- Better solution
- Still few knots

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Travelling Salesperson (5)



- After 50,000 generations
- Near-optimal solution
- No knots

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Domains of Application (5)

□ Design

- Pipe network strengthening
- Metro Toronto (Canada)
- Problem size = $9.16 \cdot 10^{847}$
- Solution cost:
 - Previous study: \$123 Mil.
 - GA study: \$110 Mil.
 - over 10% savings

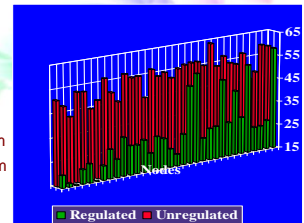


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Domains of Application (6)

□ Control

- Excess pressure minimization
- $1.4 \cdot 10^{14}$ solutions
- Solution
 - Before GA: 1257.5 m
 - GA: 473.5 m
 - over 60% reduction
- 50,000 evaluations



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Performance

- Acceptable performance at acceptable costs on a wide range of problems
- Intrinsic parallelism (robustness, fault tolerance)
- Superior to other techniques on complex problems with
 - lots of data, many free parameters
 - complex relationships between parameters
 - many (local) optima

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Advantages

- No presumptions w.r.t. problem space
- Widely applicable
- Low development & application costs
- Easy to incorporate other methods
- Solutions are interpretable (unlike NN)
- Can be run interactively, accommodate user proposed solutions
- Provide many alternative solutions

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Disadvantages

- No guarantee for optimal solution within finite time
- Weak theoretical basis
- May need parameter tuning
- Often computationally expensive, i.e. slow

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Summary

EVOLUTIONARY COMPUTATION:

- is based on biological metaphors
- has great practical potentials
- is getting popular in many fields
- yields powerful, diverse applications
- gives high performance against low costs
- **AND IT'S FUN !**

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