

# Evolutionary Computation

IEEE Computational Intelligence Society  
Pre-College Subcommittee

# Optimization

- We come across optimization problems everyday.
- Anne wants to take a graduation trip to Europe. Which route shall she choose to make the flight fares cheapest?



<https://en.wikipedia.org/wiki/Europe>

# Optimization

- Brian is an archaeologist. He found a digsite and wanted to carry the most valuable things with his bag. Which ones should he choose to fit in his knapsack?



<http://www.cs.rpi.edu/academics/courses/spring12/proglang/pa3/pa3.html>

# Optimization

- These problems are easy to solve. We can try all possible options.
- What if Anne is a postwoman and has to visit 100 different places?
- It takes her  $3 \cdot 10^{141}$  years to find the best trip if she can check a billion solutions in just one second !
- Life of the universe:  $1.5 \cdot 10^9$  yrs.

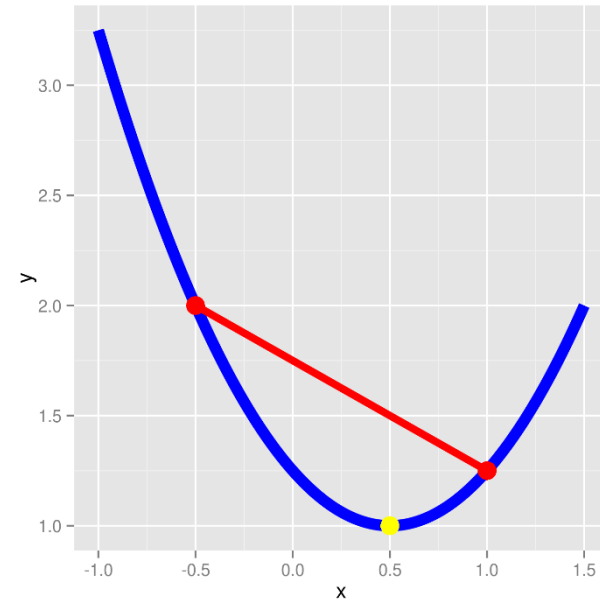
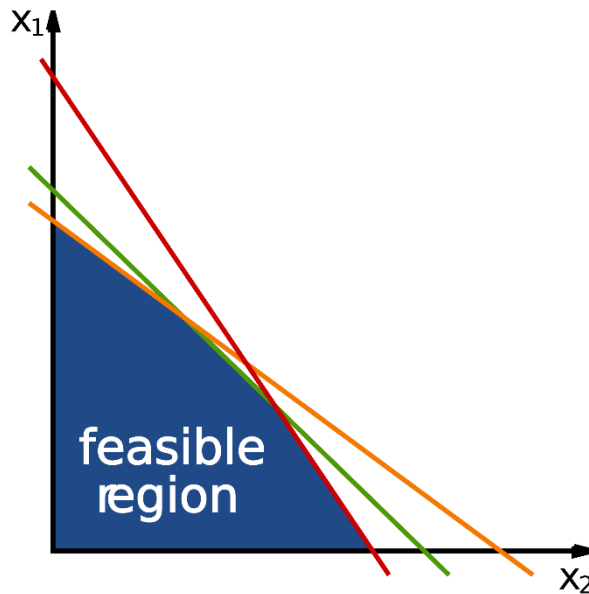
# Optimization

- We need more advanced techniques.



# Evolutionary Computation

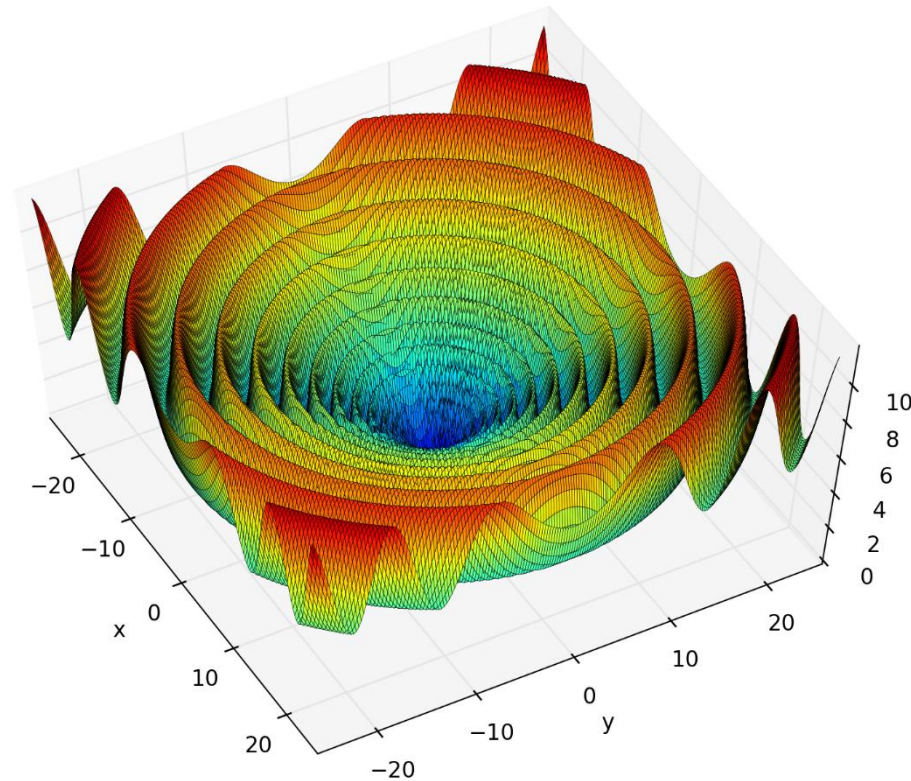
- People have developed many techniques to solve optimization problems with special characteristics.



<http://blog.ryanwalker.us/2013/09/optimization-in-r-part-i.html>

# Evolutionary Computation

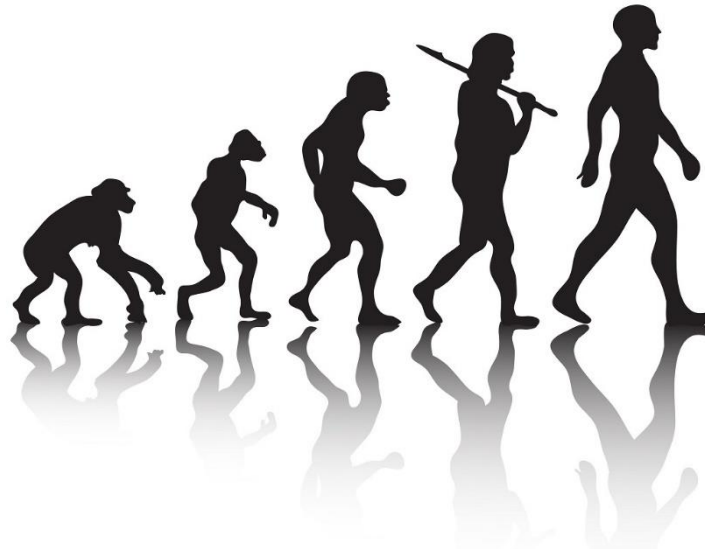
- However they cannot be used to solve too complex ones.



<http://deap.gel.ulaval.ca/doc/0.8/api/benchmarks.html>

# Evolutionary Computation

- Evolutionary computation is among the most efficient techniques to solve these problems.
- It uses the Darwinian principles for automated problem solving.





# Evolutionary Computation

- It is a kind of algorithm with the following features
  - Iterative progress
  - Population based
  - Random process
  - (Often) biological inspired

# Iterative Progress

- In evolutionary computation, the optimization problem is solved in an iterative manner.
- Each iteration tries to improve the solutions to the problem.
  - At first Anne visited all 100 places with a trip of 100km.
  - Then later she worked out another trip of 90km.
- So which solution is better ?

# Population Based

- In evolutionary computation, multiple solutions are used interacted.
  - Anne checks five different routes simultaneously.
  - In the next iteration she changed some of them according to others.

# Random Process

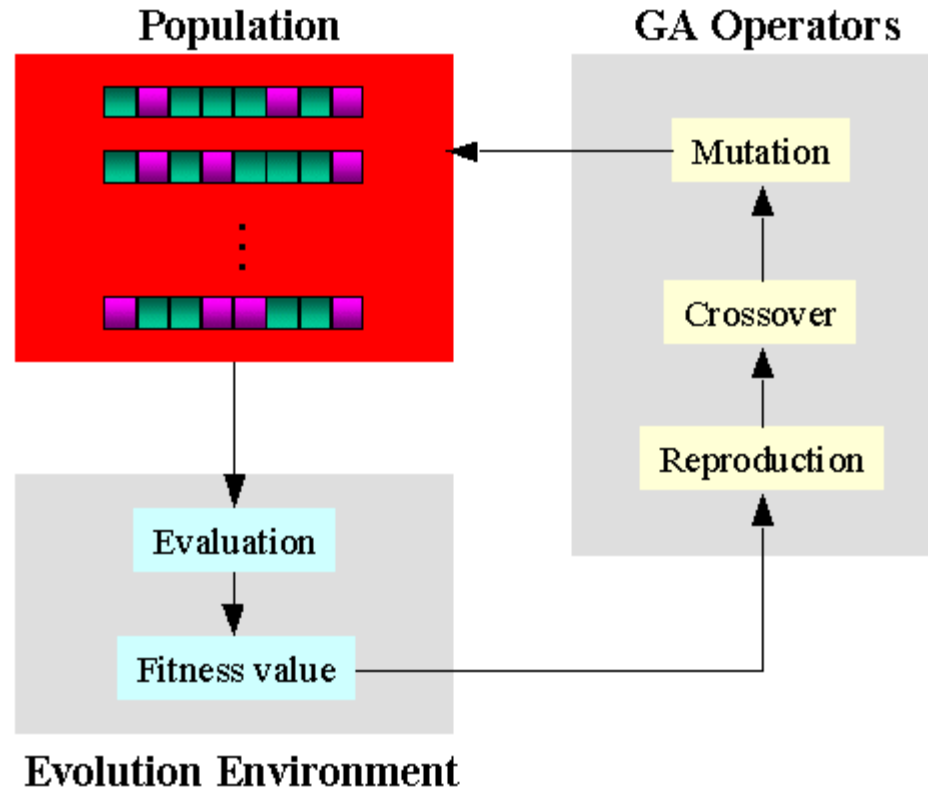
- In evolutionary computation, changes to the existing solutions are guided by a random process.
  - Anne tried to change solution A according to B. However, the resultant solution C is not constant if Anne tried another time.

# Genetic Algorithm

- Arguably the most famous evolutionary computation technique.
- Developed by John Holland in 1970s.



# Genetic Algorithm

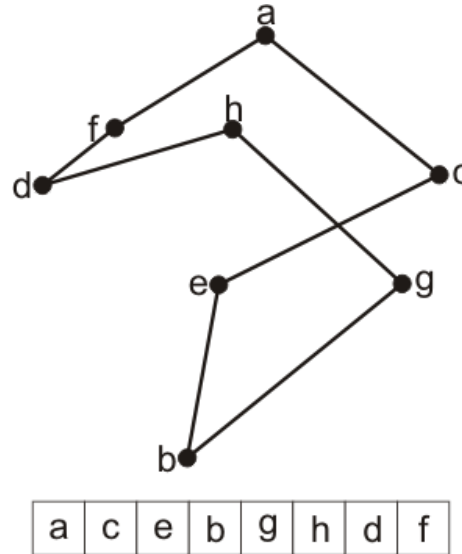


Genetic Algorithm Evolution Flow

<http://www.ewh.ieee.org/soc/es/May2001/14/Begin.htm>

# Genetic Algorithm

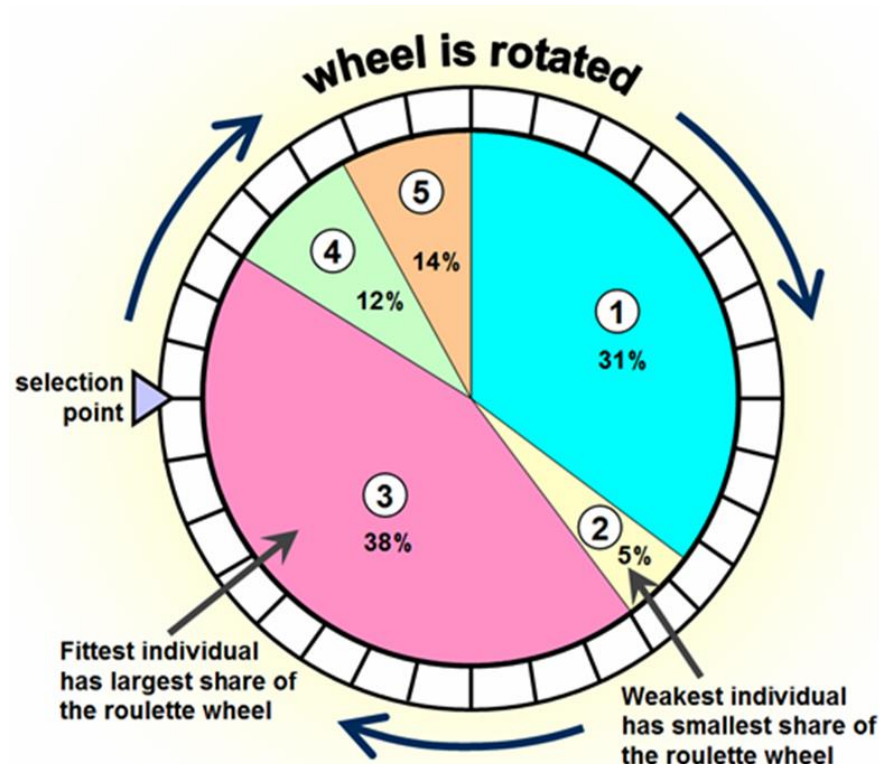
- Before optimization
  - How to represent a solution?
  - What is the objective function?
  - What is the population size?



<http://www.codeproject.com/Articles/26203/Genetic-Algorithm-Library>

# Genetic Algorithm

- Reproduction: how to select two solutions?
  - Roulette Wheel



<http://www.edc.ncl.ac.uk/highlight/rhjanuary2007g02.php>



# Genetic Algorithm

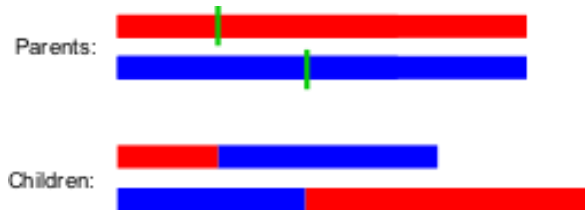
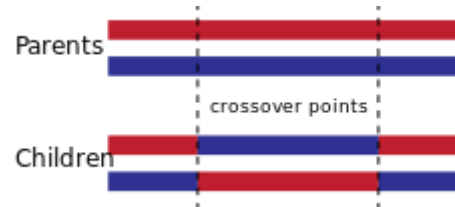
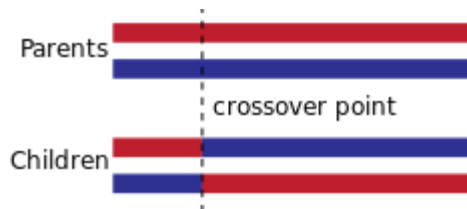
- Reproduction: how to select two solutions?
  - Tournament



[http://www.snipview.com/q/Tournament\\_selection](http://www.snipview.com/q/Tournament_selection)

# Genetic Algorithm

- Crossover: how to change two solutions?
  - Depends on solution representation

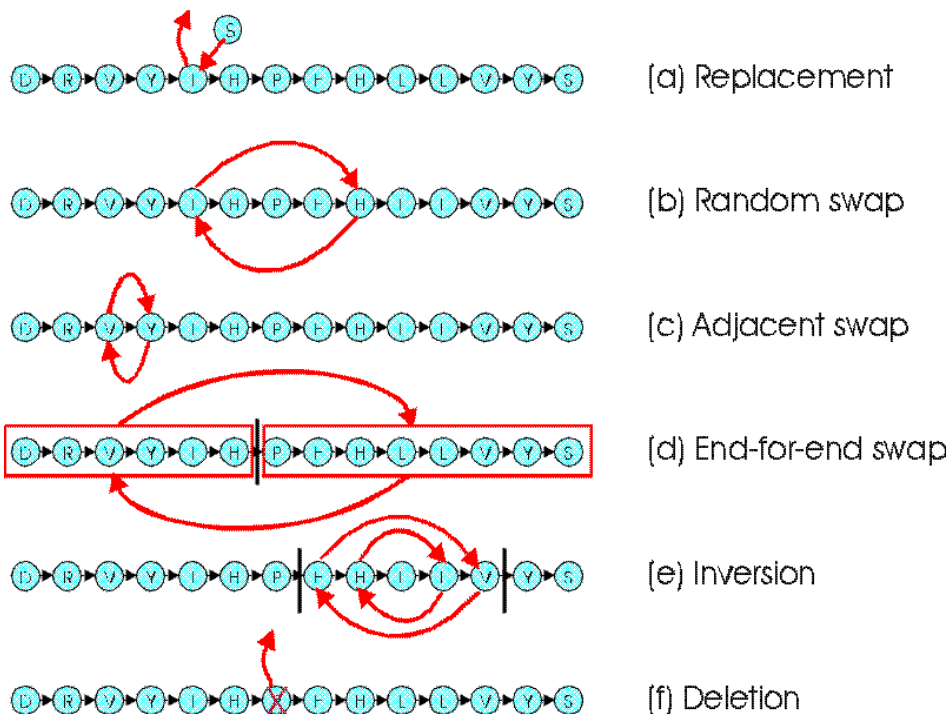


With a probability of 0.5, children have 50% genes from first parent and 50% of genes from second parent even with randomly chosen crossover points.

[https://en.wikipedia.org/wiki/Crossover\\_\(genetic\\_algorithm\)](https://en.wikipedia.org/wiki/Crossover_(genetic_algorithm))

# Genetic Algorithm

- Mutation: how to randomly change solutions?
  - Also depends on solution representation



# Genetic Algorithm

- Advantages
  - Robust
  - Simple
  - Little knowledge/assumption required
  - Wide adaptability
  - And more...
- Disadvantages
  - No guaranteed optima
  - Computationally expensive (or, slow)

# Example

- Suppose you are Brian. Now you find 7 items with different values and weights
- Item:     1       2       3       4       5       6       7
- Value:    5       8       3       2       7       9       4
- Weight:   7       8       4       10      4       6       4
- And you can carry 22 pounds of items.
- Which ones to choose?

# Example

- Before optimization
  - How to represent a solution?
    - If an item is selected, it is represented by 1's. Otherwise 0's.
    - Some possible solutions
      - 0101010
      - 1101100
      - 0100111
  - What is the objective function?
    - The total value of all selected items
  - What is the population size?
    - Reasonable sizes, e.g., 20.

# Example Solution 1

- Item:     1     2     3     4     5     6     7
  - Value:   5     8     3     2     7     9     4
  - Weight:  7     8     4    10    4     6     4
  - Select:  0     1     0     1     0     1     0
- 
- Total value:  $8 + 2 + 9 = 19$
  - Total weight:  $8 + 10 + 6 = 24 > 22 !$ 
    - It won't work!

# Example Solution 2

- Item:     1     2     3     4     5     6     7
  - Value:    5     8     3     2     7     9     4
  - Weight:   7     8     4    10    4     6     4
  - Select:   1     1     0     0     1     0     0
- 
- Total value:  $5 + 8 + 7 = 20$
  - Total weight:  $7 + 8 + 4 = 19$ 
    - Is this one the best solution?



# Example Solution 3

- Item:     1     2     3     4     5     6     7
  - Value:   5     8     3     2     7     9     4
  - Weight:  7     8     4    10    4     6     4
  - Select:  0     1     0     0     1     1     1
- 
- Total value:  $8 + 7 + 9 + 4 = 28$
  - Total weight:  $8 + 4 + 6 + 4 = 22$ 
    - This one is better!

# Example

- Putting items in a knapsack is simple.
- It can be applied to really big scenarios
  - Space ship missions have limited loads.



# More?

- IEEE Computational Intelligence Society Video Competition winning video:
  - <https://vimeo.com/76702412>

