EVOLUTIONARY ALGORITHMS

The evolutionary algorithm is a search system

- □ It handles several elements (individuals) of the problem space at each iteration and permanently modifies this population that becomes better and better until the solution (the goal, the best individual) appears.
- An individual is better than the other if it is somehow closer to the correct answer. The fitness function tries to measure this "distance" between an individual and the solution.
- □ First an initial population is selected mostly at random. In every turn the population is updated with the mutated offspring of the individuals that are selected for reproduction. This change of the population is an irreversible step, so the evolution is controlled by an irrevocable strategy.

Evolutionary operators and termination condition

- □ *Selection*: better individuals are selected for reproduction
- □ *Recombination* (*crossover*): pairs of the selected individuals (parents) are bred in order to create their offspring
- *Mutation*: offspring can be changed a bit
- *Replacement*: a new population is constructed from the elder one and the mutated offspring
- □ Termination condition:
 - either a goal individual appears in the population
 - or the overall fitness value of the population is not being changed

```
DATA := initial value

while ¬ termination condition(DATA) loop

SELECT R FROM rules that can be applied

DATA := R(DATA)

endloop

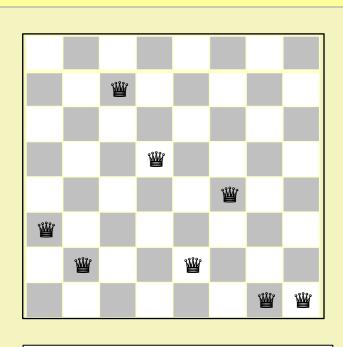
Evolutionary algorithm
```

```
Procedure EA
population := initial population
while not termination condition loop
  parents := selection(population)
  offspring:= recombination(parents)
  offspring := mutation(offspring)
  population := replacement(population, offspring)
endloop
```

Playing Super Mario with neuroevolution

- https://www.youtube.com/watch?v=qv6UVOQ0F44
- □ https://www.youtube.com/watch?v=iakFfOmanJU

n-queens problem 1

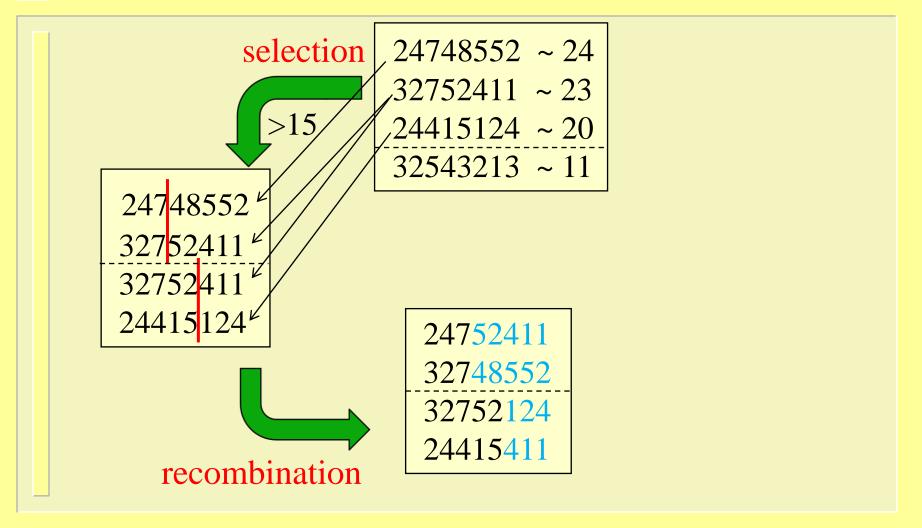


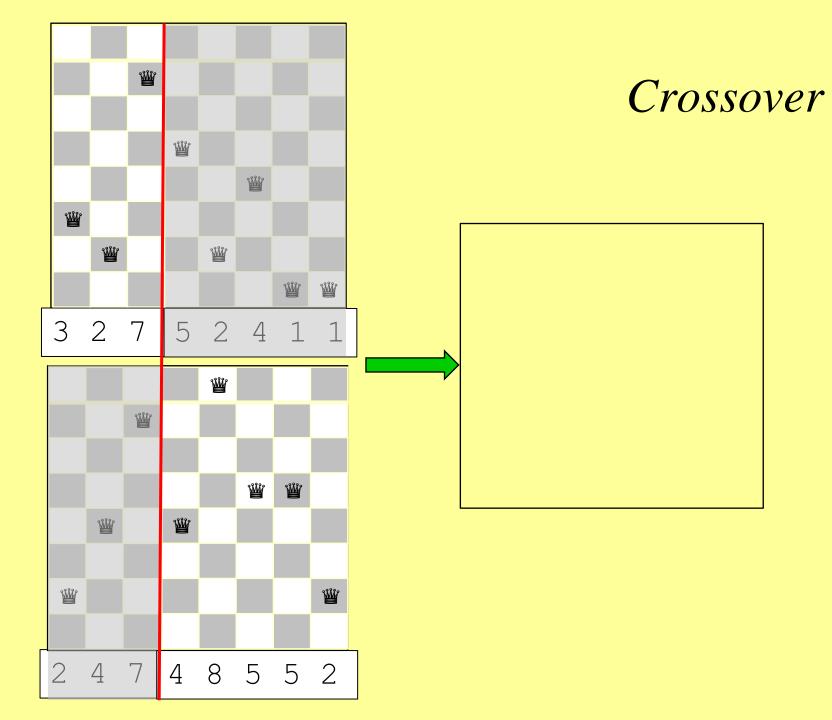
☐ Individual: a possible placement of the queens where each column contains exactly one queen

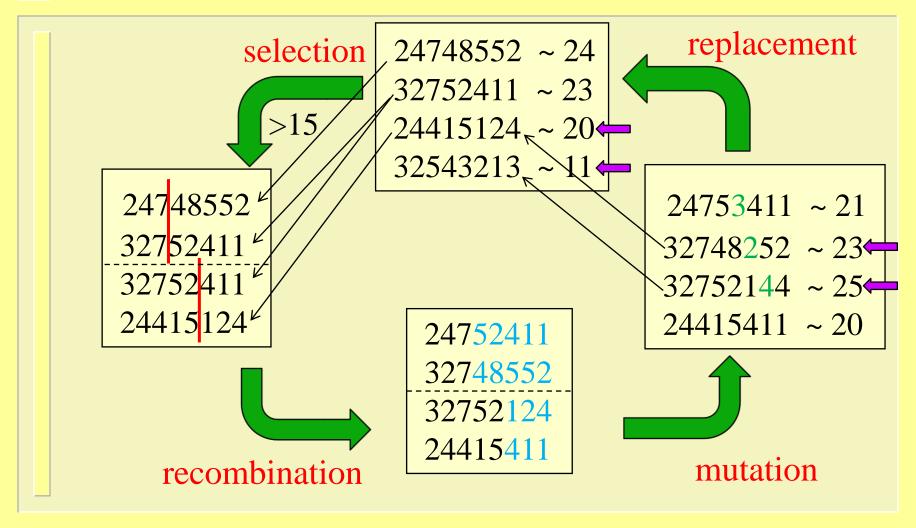
3 2 7 5 2 4 1 1

fitness value: 23

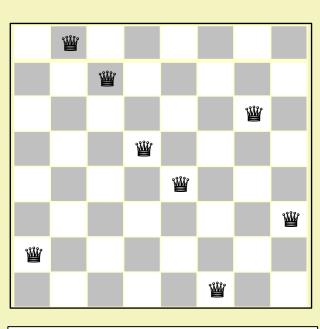
- Representation: the sequence of the row positions of the queens
- □ <u>Fitness function</u>: number of nonattacking pairs of queens







n-queens problem 2



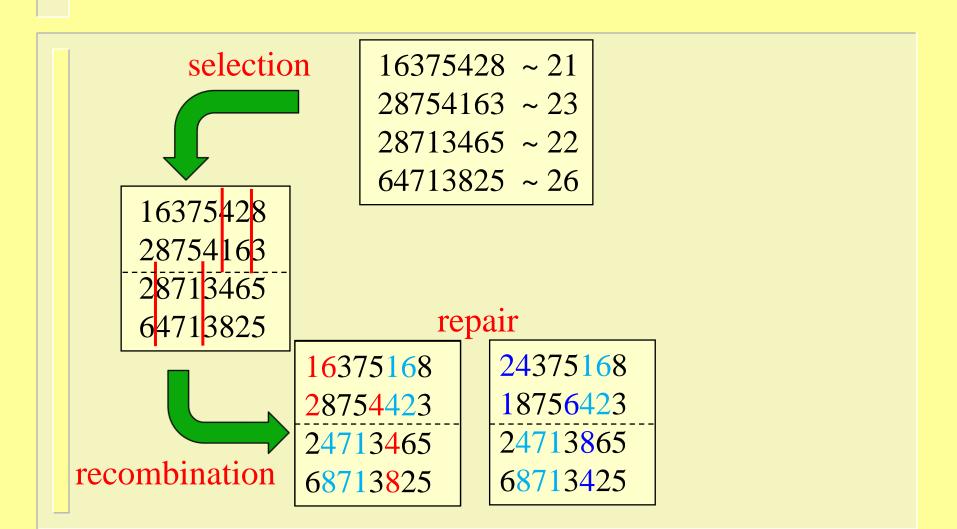
☐ Individual: a possible placement of the queens where each column and each row contains exactly one queen

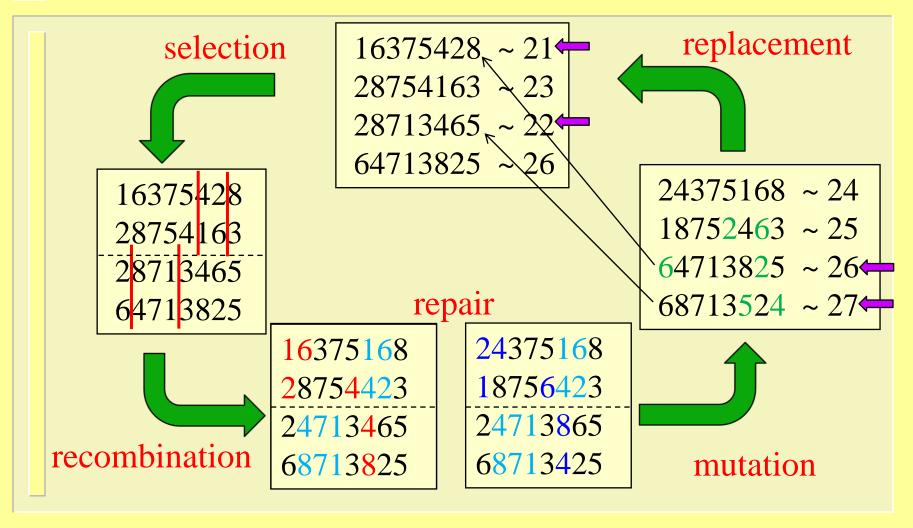
2 8 7 5 4 1 6 3

Representation: permutation of the row positions: unique (non-repeating) numbers

fitness value: 23

→ Fitness function: number of nonattacking pairs of queens





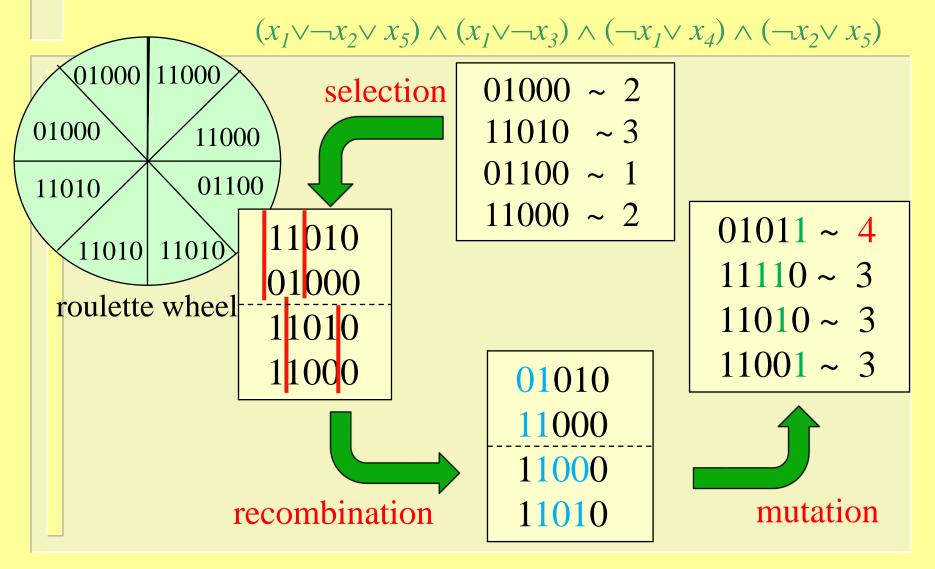
Satisfiability problem (SAT)

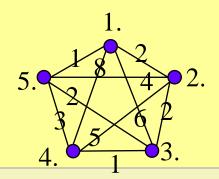
There is a Boolean statement given in CNF with n Boolean variables. Find a truth assignment for the variables so that the formula is true.

E.g.:
$$(x_1 \lor \neg x_2 \lor x_5) \land (x_1 \lor \neg x_3) \land (\neg x_1 \lor x_4) \land (\neg x_2 \lor x_5)$$

one solution: $x_1 = false, x_2 = true, x_3 = false, x_4 = true, x_5 = true$

- □ <u>Individual</u>: a possible truth assignment
- □ Representation: a sequence of bits
- □ <u>Fitness function</u>: the number of the clauses of the formula that has true value in the given truth assignment





Traveling salesman problem

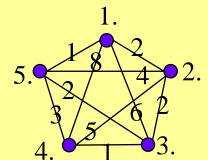
The traveling salesman must visit every city in his territory exactly once and then return to home city covering the shortest distance. (All distances between cities are known)

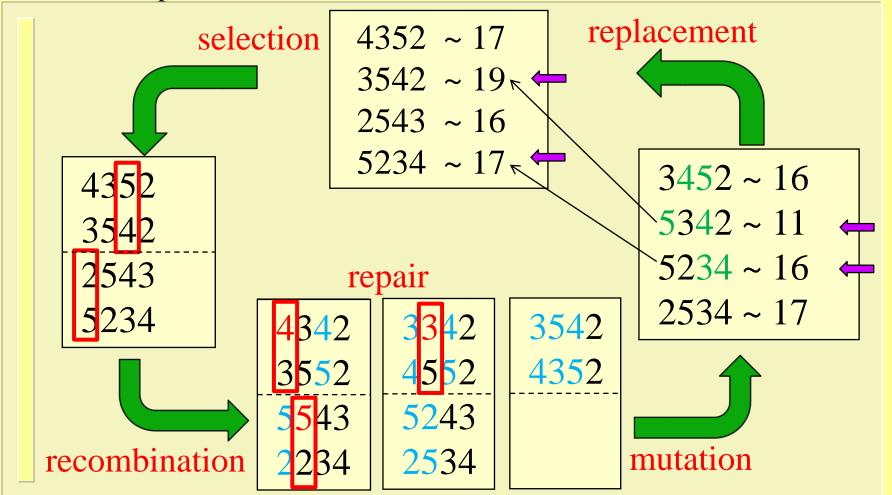
☐ <u>Individual</u>: one tour

□ Representation: permutation of the cities (numbers)

without the home city

□ Fitness function: cost of the tour





Elements of the evolutionary algorithm

- Model: individuals and their representation
- □ Fitness function (heuristics)
- Evolutionary operators
 - selection, recombination, mutation, replacement
- □ Initial population, termination condition (goal)
- Settings of the strategy parameters
 - size of the population, probability of mutation, rate
 of the offspring, rate of the replacement

Representation

- An individual is represented by a code (chromosome) that is most commonly a sequence of signals.
- □ One signal or one section of signals with its position in the code (gene) can describe one property of the individual.
 - Thus the structure of the code can be dismembered property by property. If one signal is changed, then one property of the individual is changed a bit, too.
 - Frequent coding methods:
 - Array: fixed length sequence of real numbers or integers
 - Binary code: fixed length sequence of bits
 - Permutation of the elements of a finite set

Representations and fitness functions of graph coloring problem

Color the vertices of a simple finite undirected graph with four colors so that no two adjacent vertices share the same color

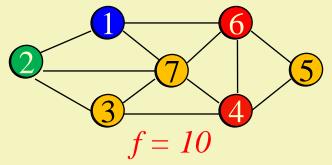
Direct encoding



1. 2. 3. 4. 5. 6. 7. 2

x[i] is the color of the ith vertex.

f is the number of the correct edges.

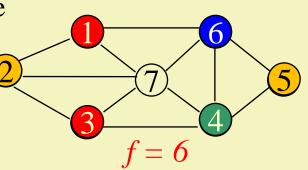


Indirect encoding

2 5 1 3 4 6 7

In the i^{th} step of the coloring process the $x[i]^{th}$ vertex must be colored with the possible lightest color regarding the colors of neighbor vertices, if possible.

f is the number of the colored vertices.



Representation and fitness functions of problem of rock-paper-scissors

Find a good strategy for the rock-paper-scissors world championship!

- □ A strategy should give our next step based on a few (say two) previous games.
 - For example:

Antecedent: Me: PR Suggestion: S

Op: S P

• This is only one slice of the strategy. We need a suggestion for each possible antecedent.

Representation

There are 381 different strategies (individuals) and their codes are the elements of $\{0,1,2\}^{0..80}$

Signals

 $R \sim 0$

Antecedents (MeOpMeOp)

RRRR ~ 0000 ~ 0

RRRP ~ 0001 ~ 1

RRRS ~ 0002 ~ 2

RRPR ~ 0010 ~ 3

SSSP ~ 2221 ~ 79

SSSS ~ 2222 ~ 80

Suggestions

Code of the strategy: 1201 ... 20

Fitness function

```
Strategy: 1201 ... 20
```

Specimen battle:

Player: 0002221222001000

Op: 0102112220101011

Signal

 $R \sim C$

P ~ 1

S ~ 2

Selection

- □ The better individuals must be selected but the worse ones must be given a chance to be chosen. (stochastic method)
 - Roulette wheel: the probability of selection of an individual is proportional to its fitness function value
 - Ranking: the probability of selection of an individual is proportional to its ranking based on the fitness function values
 - Tournament: the selected individuals are the best individuals of randomly selected groups of the population
 - Culling: all individuals below a given threshold are discarded and then the individuals are selected randomly from the remaining individuals

Recombination

- □ Offspring must be created from the pairs of parents so that the offspring inherit the properties of their parents.
 - Crossover: signals of the parent codes are switched at the positions chosen randomly
 - Recombination: the corresponding signals of the parent codes are combined
- ☐ If the code has got some extra invariant condition (permutation), it must be preserved.

Crossover

□ Single or multiple pointed crossover

 Some crossover points are chosen at random and the signals are exchanged in every second section.

□ Uniform crossover

 Some positions are chosen at random and only the signals on these positions are exchanged.

Crossover for permutations 1

□ Partial crossover

 Signals are exchanged in one section chosen at random and then the duplicated signals of the first and the second parent besides the crossover section are paired and switched.

$$\begin{bmatrix}
2 & 3 & 1 & 5 & 4 & 6 & 7 \\
1 & 7 & 4 & 2 & 5 & 3 & 6
\end{bmatrix}
\longrightarrow
\begin{bmatrix}
2 & 7 & 4 & 2 & 4 & 6 & 7 \\
1 & 3 & 1 & 5 & 5 & 3 & 6
\end{bmatrix}
\longrightarrow
\begin{bmatrix}
1 & 7 & 4 & 2 & 5 & 6 & 3 \\
2 & 3 & 1 & 5 & 4 & 7 & 6
\end{bmatrix}$$

Crossover for permutations 2

□ Cycle crossover

- 1. Select $i \in [1..length]$ at random
- 2. Change $x_i \leftrightarrow y_i$
- 3. Find $j \in [1..length]$ $(j \neq i)$ where $x_j = x_i$,
- 4. If it is not found then over else i = j
- 5. goto 2.

Recombinations of real vectors

□ General recombination

- The offspring (\underline{u}) is somewhere around or in the hyper cube spanned by the parents (a, b)
- $\forall i$ ∈1... n : $u_i = c_i \cdot a_i + (1 c_i) \cdot b_i$ where c_i ∈[-h, 1+h] selected randomly

□ Linear recombination

- The offspring (\underline{u}) is somewhere on the line matching the parents (a, b)
- $\forall i \in 1... n : u_i = c \cdot a_i + (1-c) \cdot b_i$ where $c \in [-h, 1+h]$ selected randomly

Mutation

- \square Each position of the code is subject to random change with a small independent probability (p).
- When the code is an array including real numbers:
 - $\forall i=1...n: z_i = x_i \pm domain_i \cdot p$
- When the code is a binary code:
 - $\forall i=1...n: z_i = 1-x_i \text{ if } random[0..1] < p$
- When the code is a permutation:
 - swapping signal-pairs
 - shifting the signals in a section cyclically
 - reversing or reordering the signals in a section

Replacement

Creation of the new population from the elder one and the mutated offspring: the selected individuals of the population are substituted with the selected offspring two selections are needed

```
offspring rate(u) = \frac{\text{number of the offspring to be put in}}{\text{size of the population}}
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replacement rate(v) = $\frac{\text{number of the individuals to be removed}}{\text{size of the population}}$

- u=v: total exchange
 additional selection is needed
- u<v: some selected offspring must be used several times
 additional selection is needed
- u>v : some selected offspring must be eliminated