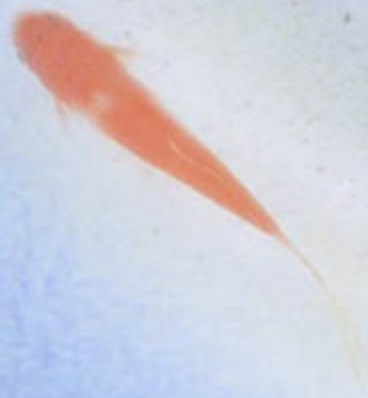


# Optimization algorithms inspired by nature

Mr.sc. Marko Čupić  
Zagreb, June 2., 2011.



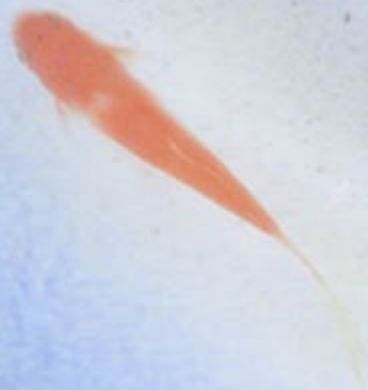
# Introduction

- Emerging intelligence
- Optimization problems
- Genetic algorithm
- Ant colony optimization

# Emerging intelligence

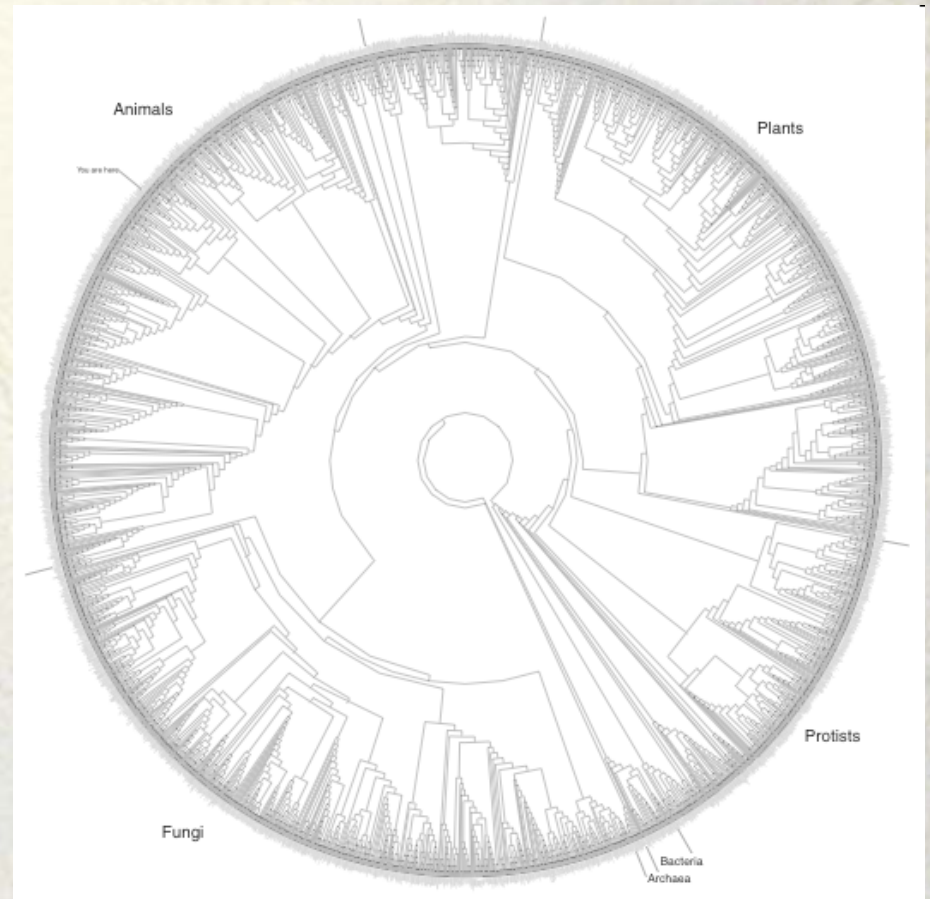
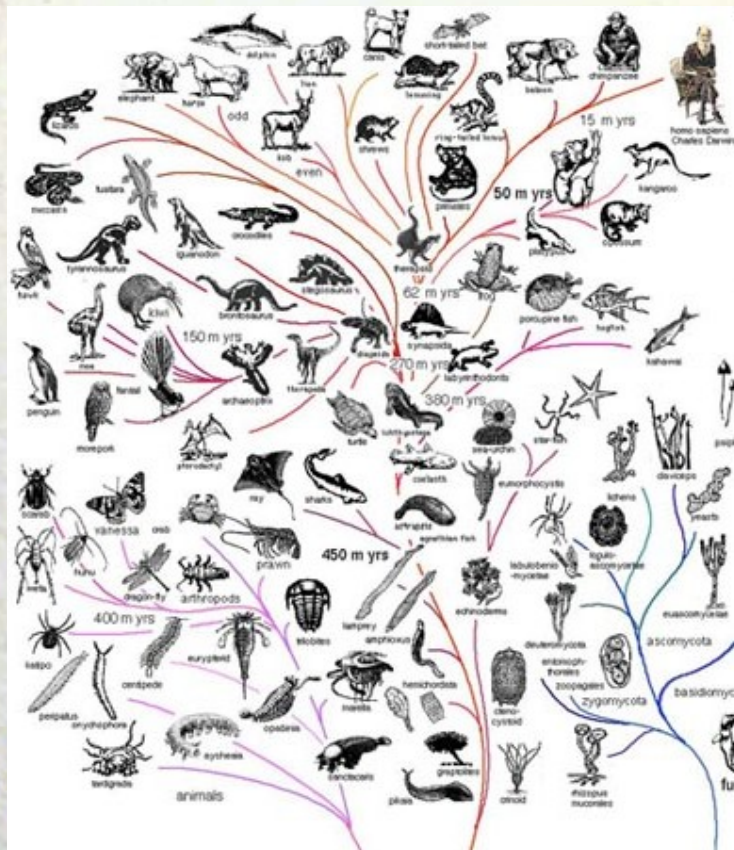
- **Emerging intelligence**

- Optimization problems
- Genetic algorithm
- Ant colony optimization



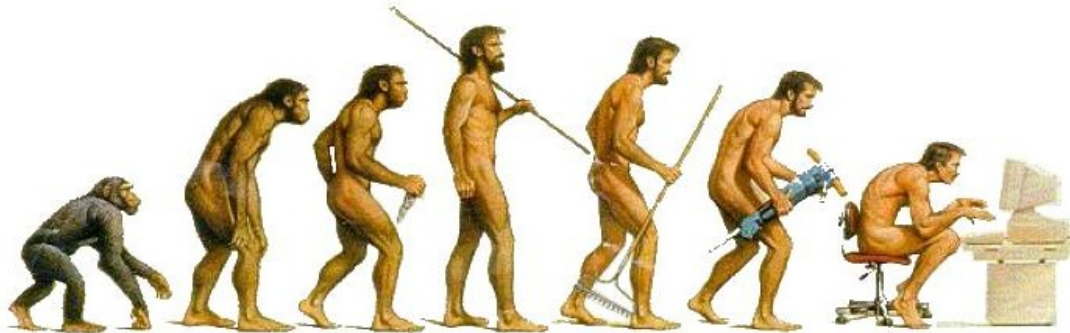


# Introduction: Evolution



# Introduction: Evolution

Evolution



(OR is it?)



# Introduction: Flock of birds





# Introduction: Bees swarm





# Introduction: Ant colonies



- Area of 50 m<sup>2</sup>, depth of 8m



# Introduction: Ant colonies



- Ant mound, path...



# Introduction: Ant colonies



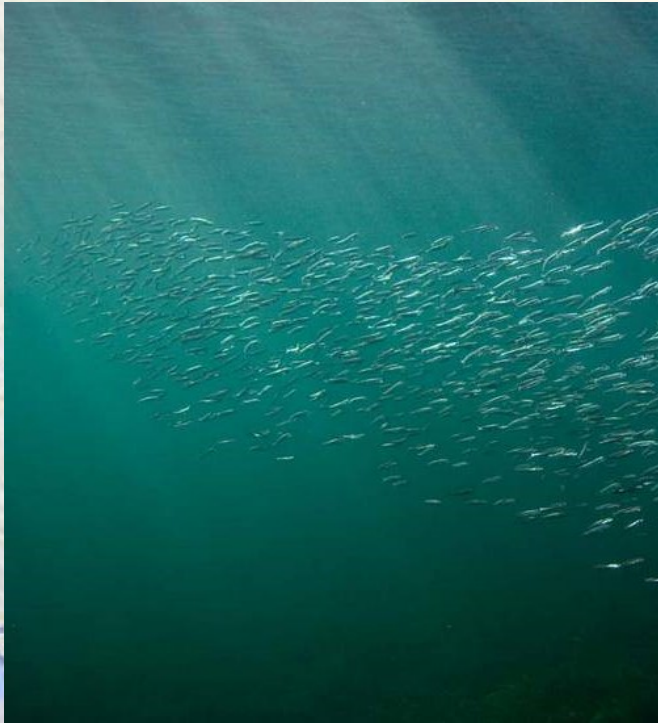


# Introduction: Ant colonies





# Introduction: School of fish





Introduction: conclusions

Dumb parts, properly  
connected  
into a swarm, yield smart  
results.

*Kevin Kelly*

*New Rules for the New Economy*

*Sep 1997*

Introduction: conclusions

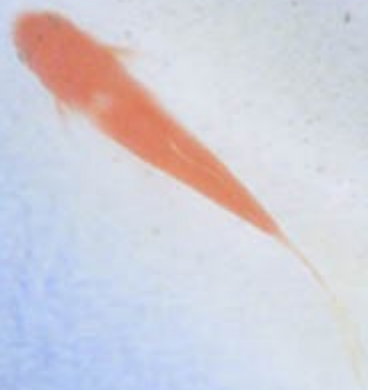
The **whole** is **greater**  
than

**the sum of the parts.**



# Optimization problems

- Emerging intelligence
- **Optimization problems**
- Genetic algorithm
- Ant colony optimization



# Optimization problems

- Optimization: the procedure of finding the best solution of a problem, the solution with the lowest price
- Typically:
  - Continuous variables
  - Combinatorial problems



# Optimization problems

- State space search:
  - Find a path from  $S_0$  to  $S_F$
  - The solution is the **path** (e.g. 3x3 jigsaw puzzle!)
- CSP: *Constraint Satisfaction Problem*
  - A kind of state space search where the path from  $S_0$  to  $S_F$  is not important. The solution is the final state itself.

# Optimization problems

- CSP: *Constraint Satisfaction Problem*
  - Constraints that must be satisfied are defined
  - A criterion function to optimize is given



# Optimization problems

- We've seen ways to tackle combinatorial problems
  - State space search algorithms
    - Breadth first search
    - Depth first search
    - $A^*$
    - ...
- Unfortunately, often not applicable to real problems

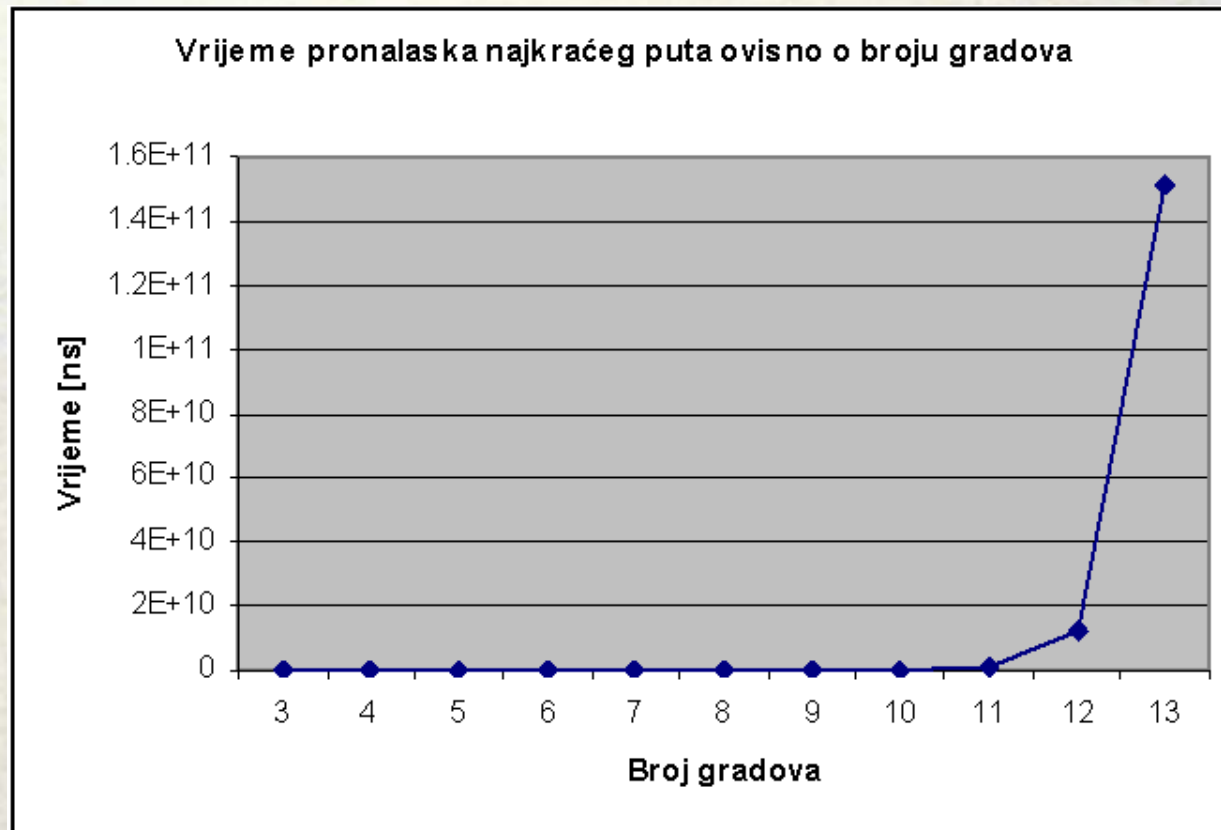
# Optimization problems

- The traveling salesman problem
  - Coordinates of  $n$  cities on the map are given
  - Find the shortest tour through all cities
  - Mathematically: find the shortest Hamiltonian cycle in the graph
  - **NP hard problem**  
(factorial complexity)



# Optimization problems

- Traveling salesman problem



# Optimization problems

- Traveling salesman problem
  - 12 cities, 12 sec
  - 13 cities, 2.5 min
  - 14 cities, 0.5h
  - 15 cities, 7.6h
  - 16 cities, 4.7 days
  - ...
  - 500 cities, ?????



# Optimization problems

- Other problems
  - Scheduling unscheduled students into groups for classes
  - Midterm timetable creation
  - Lab assignments timetable creation
- Enumerating all possibilities?
  - It would take much much more time than the age of the universe

# Optimization problems

- Heuristics
  - Algorithms that find *sufficiently* good solutions, usually do not guarantee optimality, and have low computational complexity (polynomial)
  - They can be
    - Construction based
    - Local search based



# Optimization problems

- Heuristics
  - Construction based
    - Build the solution incrementally
  - Local search algorithms
    - Start with a completed solution and try to incrementally make it better

# Optimization problems

- Metaheuristics
  - A set of algorithmic concepts used to define heuristic methods applicable to a wide set of problems
  - A heuristic guiding problem specific heuristics



# Optimization problems

- Metaheuristics
  - Simulated annealing
  - Taboo search
  - Evolutionary algorithms
  - Ant colony optimization
  - Swarm optimization
  - Artificial immunological systems
  - ...

# A problem ☺

- "No free lunch" Theorem, Wolpert & Macready, 1995, 1997:
  - *All algorithms seeking an optimum of a goal function behave identically with respect to any performance measure, when considered averaged over all possible goal functions*
  - ...

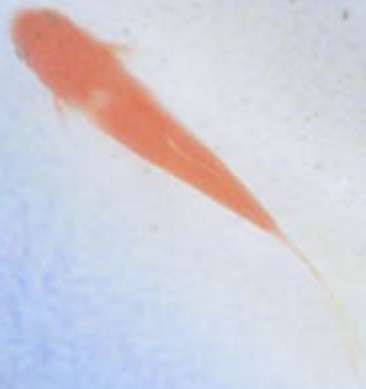


# A problem ☺

- "No free lunch" Theorem, Wolpert & Macready, 1995, 1997:
  - *Specifically, if algorithm A is better than algorithm B on some goal functions, then, roughly speaking, there must be exactly that many different goal functions on which B is better than A.*

# Genetic algorithm

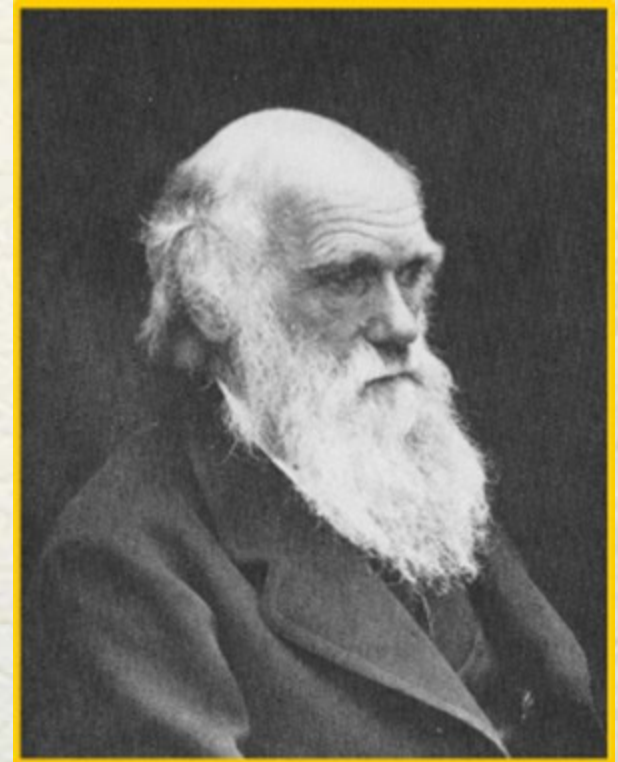
- Emerging intelligence
- Optimization problems
- **Genetic algorithm**
- Ant colony optimization





# Genetic algorithm

- Evolution as inspiration
- Population based algorithms
- Darwins theory about the origin of species



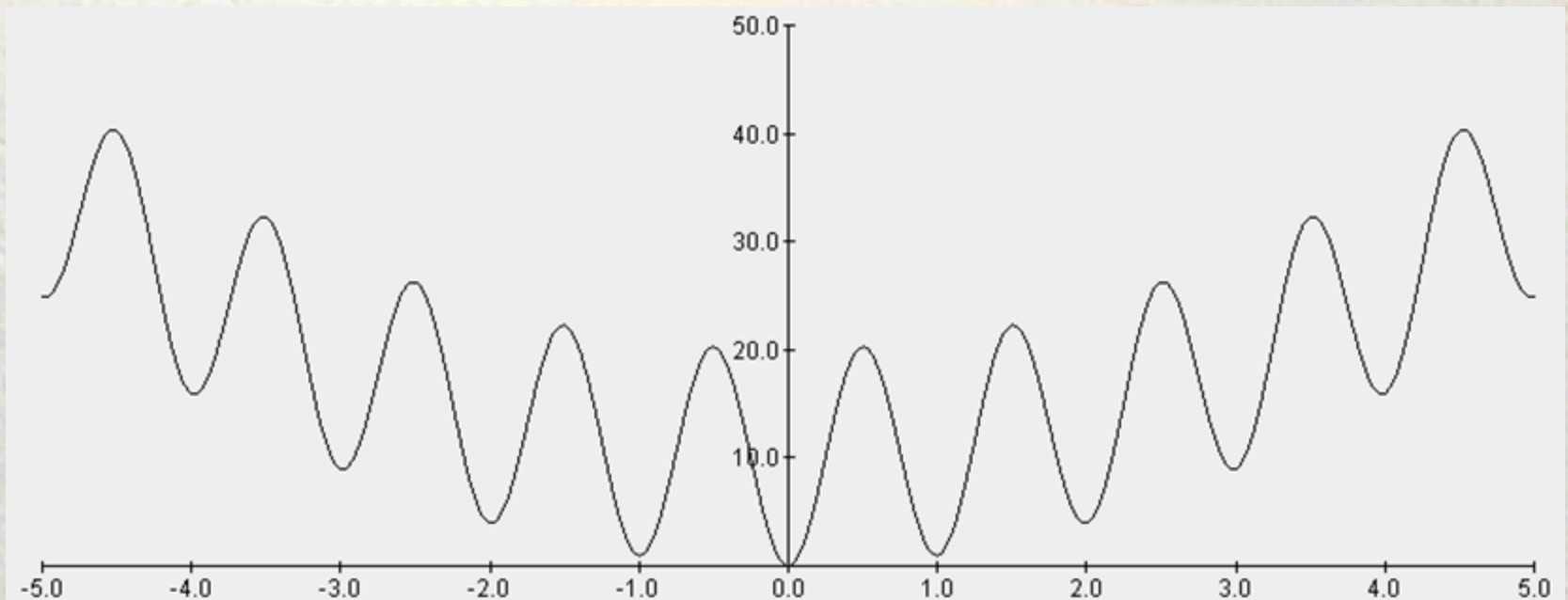
# Genetic algorithm

- Main settings: Darwin
  - Fertility of species – there are always more descendants than required
  - Size of the population is roughly constant
  - Food supply is limited
  - For species that reproduce sexually there are no identical individuals, there are variations
  - Most of an individual's specific variations is passed on to its descendants



# Genetic algorithm

- Example problem  $f(x) = 10 + x^2 - 10 \cdot \cos(2 \cdot \pi \cdot x)$ 
  - Find  $x$  for which  $f(x)$  is minimal



# Genetic algorithm

- How does GA work?
  - There is a population of chromosomes
  - Each chromosome represents one solution to the problem
  - Each solution has a fitness
  - In our example fitness and  $f(x)$  are opposite → higher  $f(x)$  means lower fitness

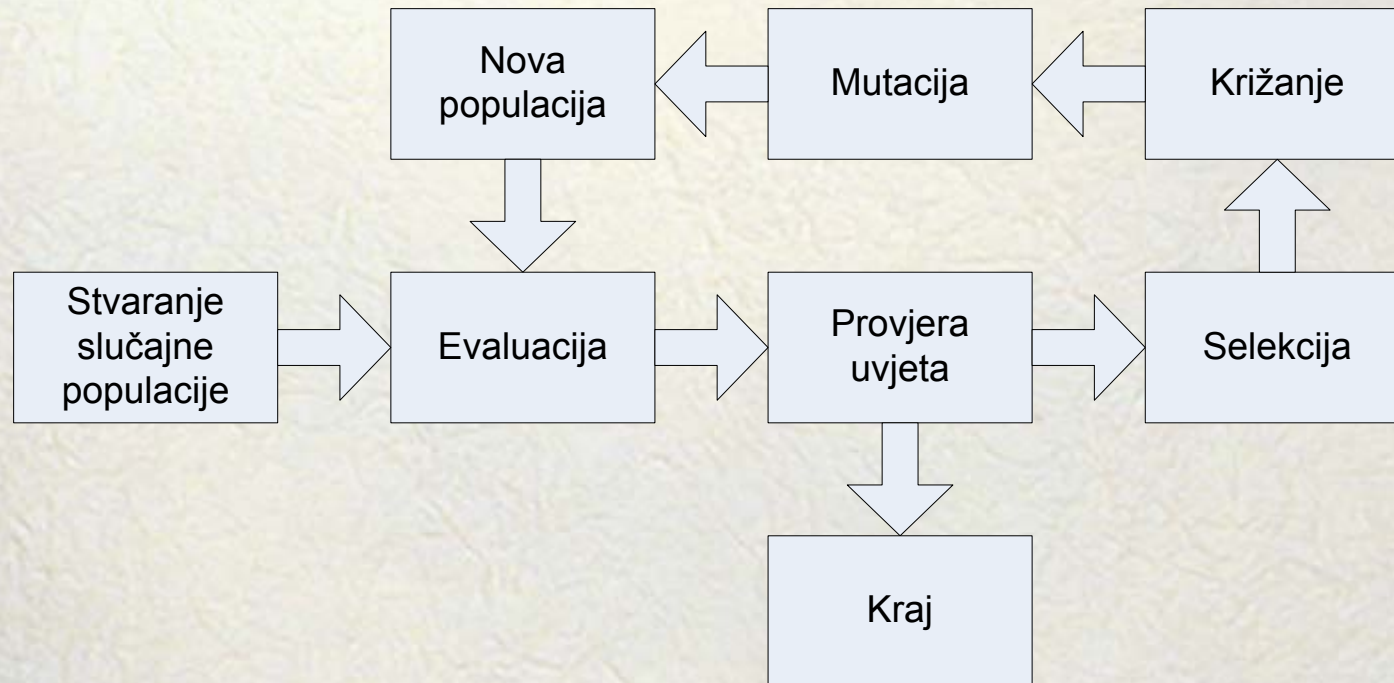


# Genetic algorithm

- Implementation
  - From the current generation we iteratively create the next one
  - We select individuals that have a higher probability of creating better solutions
  - They are combined using the crossover operator
  - Resulting individuals are mutated using the mutation operator

# Genetic algorithm

- Flowchart





# Genetic algorithm

- Roles

- **Selection** → selectional pressure → speed of convergence
- **Crossover** → searching the neighbourhood of parents
- **Mutation** → getting out of local optima, big jumps in the solution space

# Genetic algorithm

- Binary chromosome 

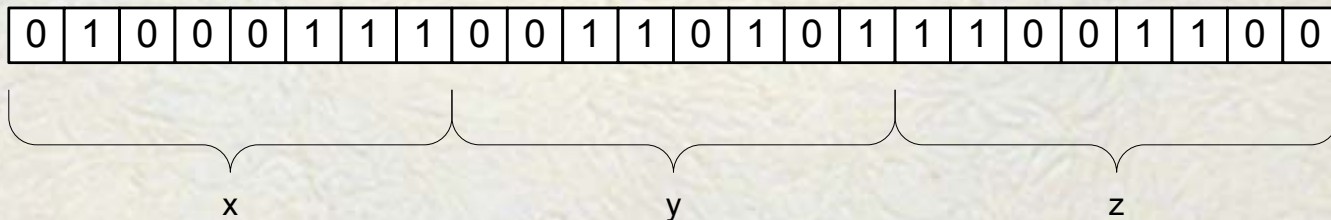
0	1	0	0	0	1	1	1
---	---	---	---	---	---	---	---

  - A sequence of binary digits interpreted as a solution (value of a variable)
  - Three bit chromosome: 000, 001, ..., 111
  - Assuming we are observing a real variable from the interval  $[-2, 2]$ , then:  
000 $\equiv$ -2, 001 $\equiv$ -1.43, ..., 111 $\equiv$ 2
  - What is the number of bits for a given precision?



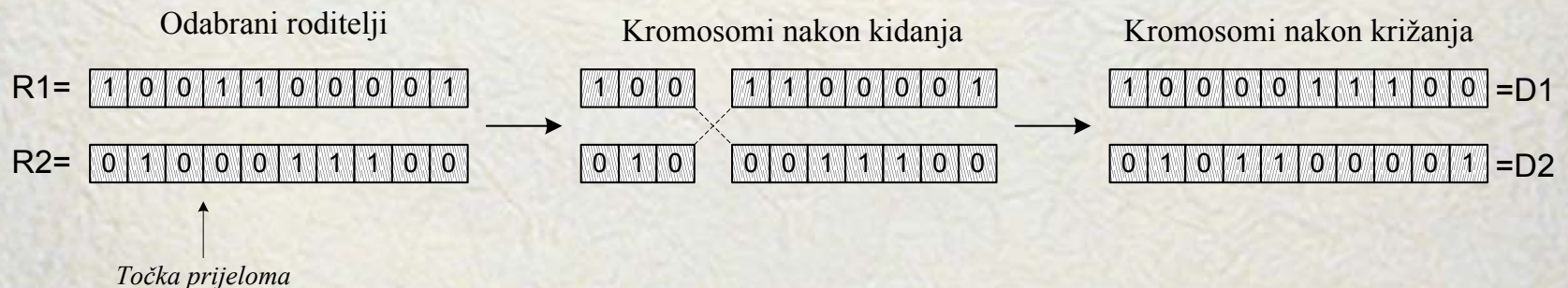
# Genetic algorithm

- Binary chromosome
  - A more complex example
    - Solution for a function of three variables  $x$ ,  $y$ ,  $z$



# Genetic algorithm

- Crossover with one breaking point
  - Two parents are chosen
  - A breaking point is randomly chosen
  - Crossover is performed



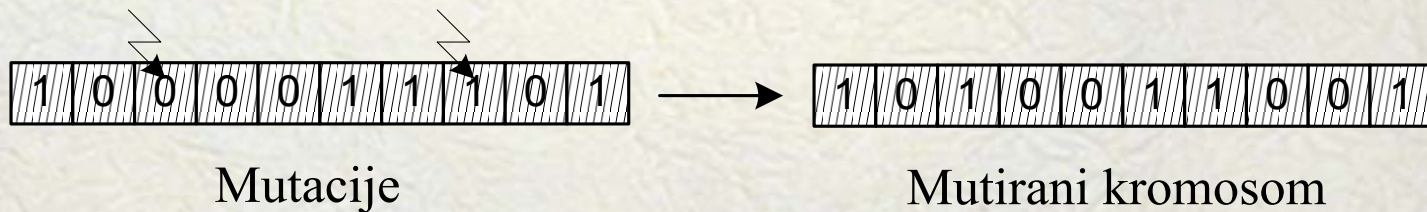


# Genetic algorithm

- Other types of crossover
  - Crossover with one breaking point
  - Crossover with  $n$  breaking points
  - Uniform crossover
  - ...

# Genetic algorithm

- Mutation operator
  - Mutation probability is given
  - Each bit is flipped with that probability



- Can introduce a huge change!



# Genetic algorithm

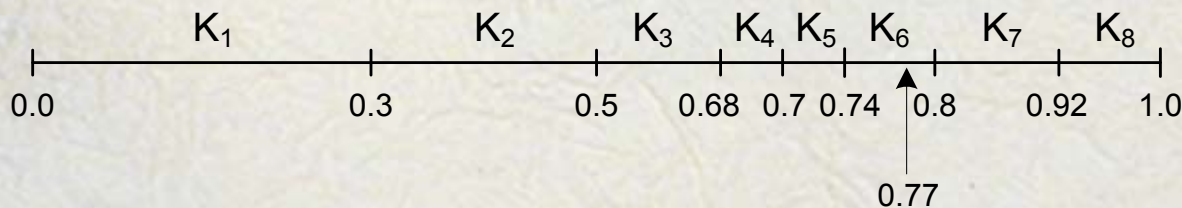
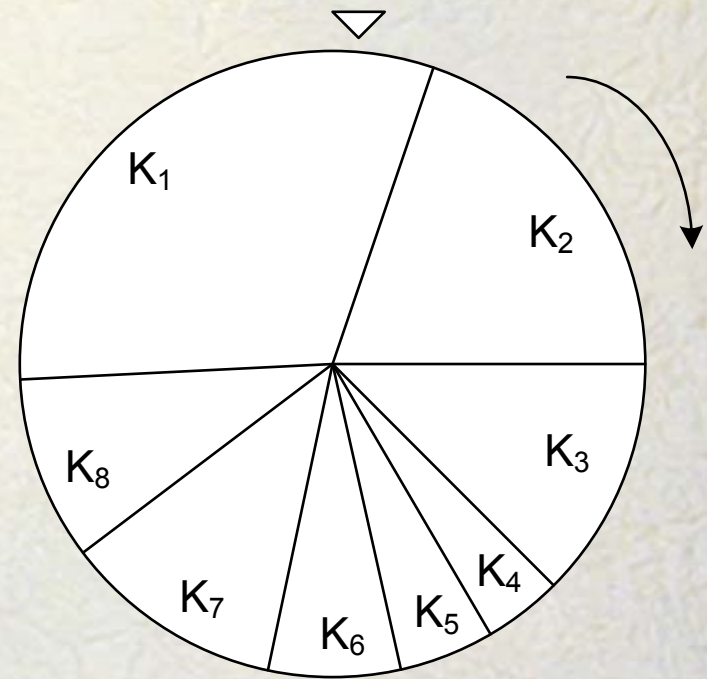
- Selection of parents
  - Proportional selection - Roulette-wheel selection
  - More fitness of an individual means higher chances in the selection process

$$probSel(i) = \frac{fit(i)}{\sum_{j=1}^n fit(j)}$$

# Genetic algorithm

- Selection of parents – proportional selection

$$len(i) = \frac{fit(i)}{\sum_{j=1}^n fit(j)}$$

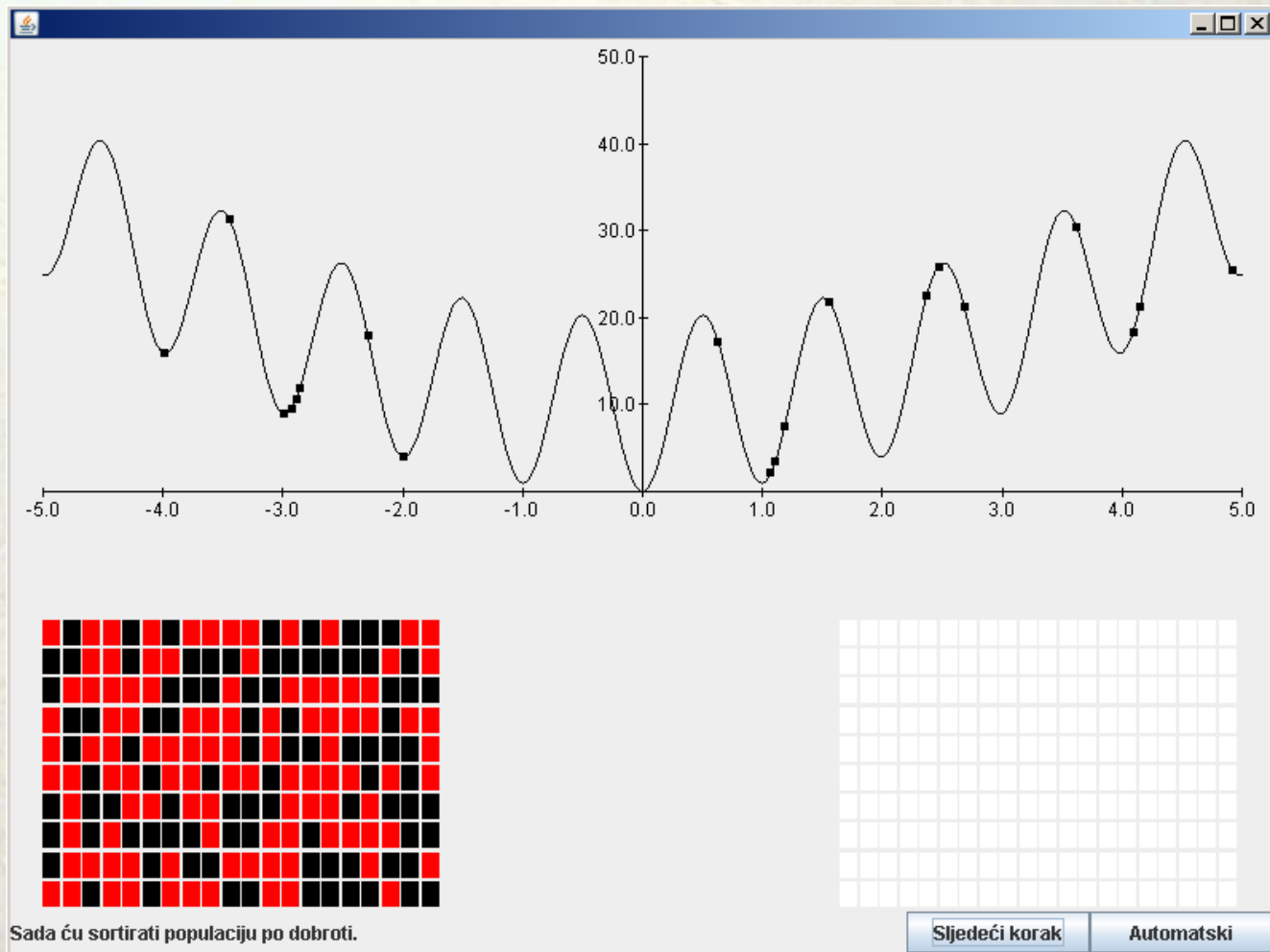




# Genetic algorithm

```
P = create_initial_population(POP_SIZE)
evaluate(P)
repeat_until_done:
  new_population P' =  $\emptyset$ 
  repeat_while size(P') < POP_SIZE
    select R1 and R2 from P
    {D1, D2} = crossover(R1, R2)
    mutate D1, mutate D2
    add D1 and D2 into P'
  end_repeat
  P = P'
  evaluate(P)
end_repeat
```

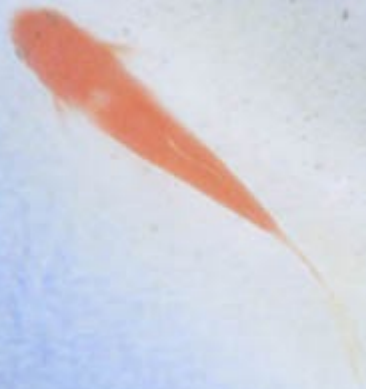
# Genetic algorithm





# Ant colony optimization

- Emerging intelligence
- Optimization problems
- Genetic algorithm
- **Ant colony optimization**



# Ant colony optimization

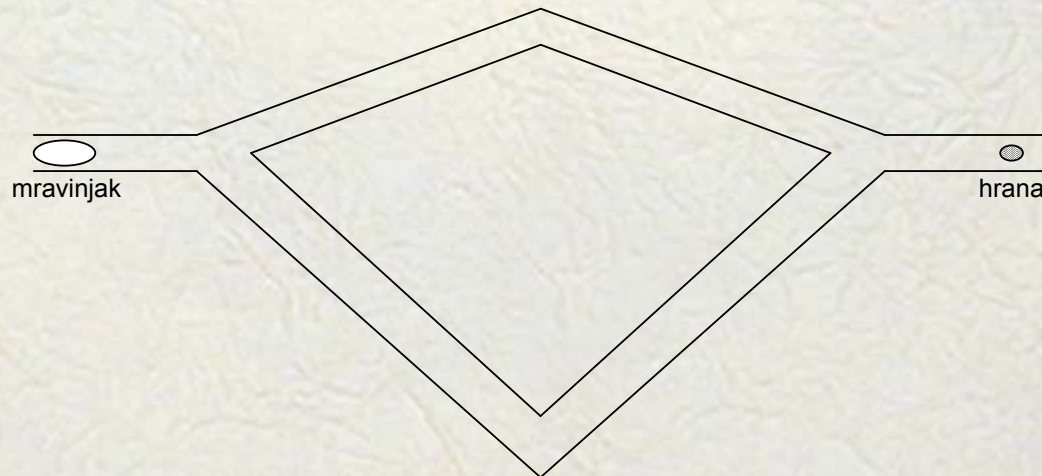
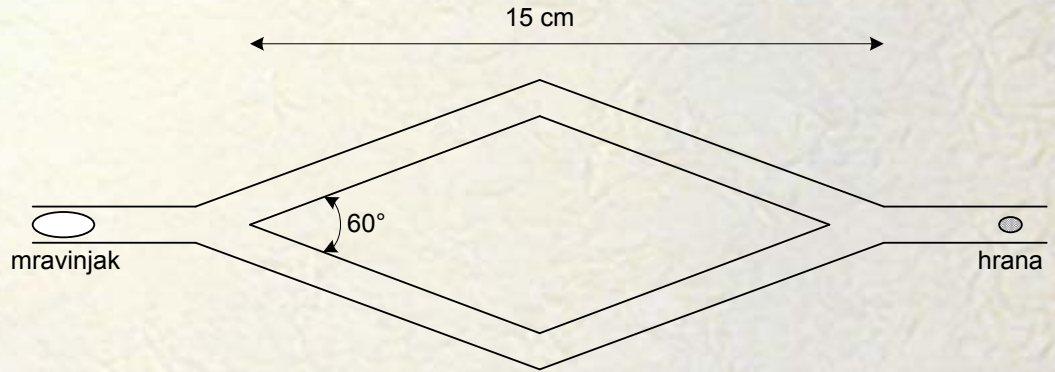
- Ants exhibit interesting behavior
  - They successfully find the shortest path to food sources





# Ant colony optimization

- Experiments



# Ant colony optimization

- Explanation
  - While moving ants leave a feromon trail behind
  - An ant moves randomly, but it is more likely to go in the direction where the feromon trail is stronger

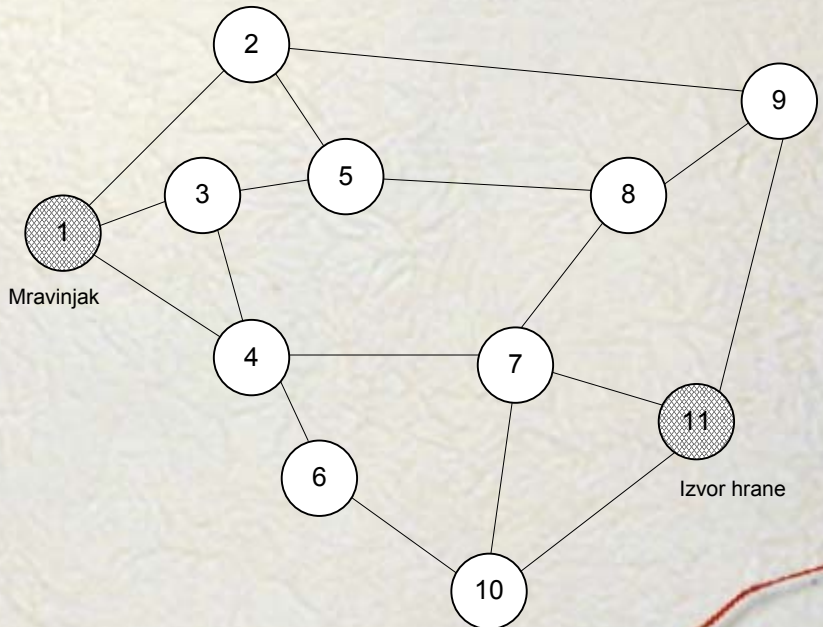


# Ant colony optimization

- Directly applicable to problems described by graphs
- E.g. From 1, possible next are 2,3,4

$$\tau_0 = konst$$

$$p_{ij}^k = \begin{cases} \frac{\tau_{ij}^\alpha}{\sum_{l \in N_i^k} \tau_{il}^\alpha} & \text{if } j \in N_i^k \\ 0, & \text{if } j \notin N_i^k \end{cases}$$

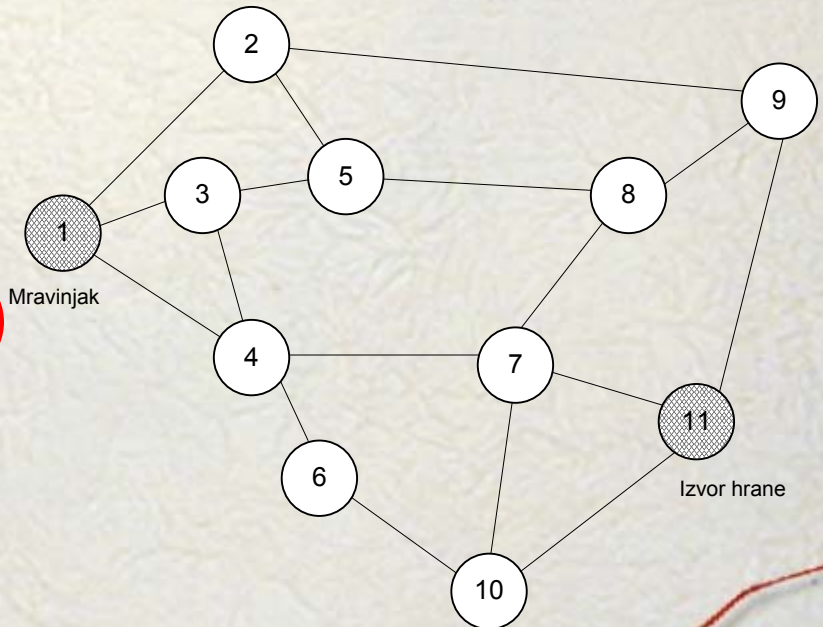


# Ant colony optimization

- Ant System algorithm
  - Using heuristic information has additional benefits on performance

$$\tau_0 = \frac{m}{C^{nn}}$$

$$p_{ij}^k = \begin{cases} \frac{\tau_{ij}^\alpha \cdot \eta_{ij}^\beta}{\sum_{l \in N_i^k} (\tau_{il}^\alpha \cdot \eta_{il}^\beta)}, & \text{ako } j \in N_i^k \\ 0, & \text{ako } j \notin N_i^k \end{cases}$$





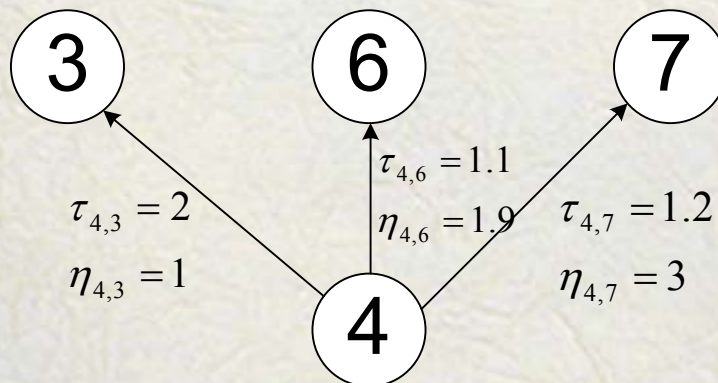
# Ant colony optimization

- Ant System algorithm

```
repeat until not done
  repeat for each ant
    create solution
    evaluate solution
  end repeat
  evaporate_pheromons
  repeat for all_or_some ants
    update_pheromons
  end repeat
end repeat
```

# Ant colony optimization

- Procedure: `Create_solution`
  - An ant starts from a node
  - With respect to the probabilities, a next node is chosen, then another, and so on until the ant reaches the last node



Uz  $\alpha=1$ ,  $\beta=2$ :

$$p(4 \rightarrow 3) = 11,9\%$$

$$p(4 \rightarrow 6) = 23,7\%$$

$$p(4 \rightarrow 7) = 64,4\%$$



# Ant colony optimization

- Procedure: Evaluate\_solution
  - Calculates the total path length
  - Moving from one node to another is usually associated with a cost (cities → distance)

# Ant colony optimization

- Procedure: Evaporate\_pheromons
  - Lowers pheromon trails on all edges by an amount

$$\tau_{ij} \leftarrow \tau_{ij} \cdot (1 - \rho)$$

- Geometric progression!
- Very costly(graph has many edges)



# Ant colony optimization

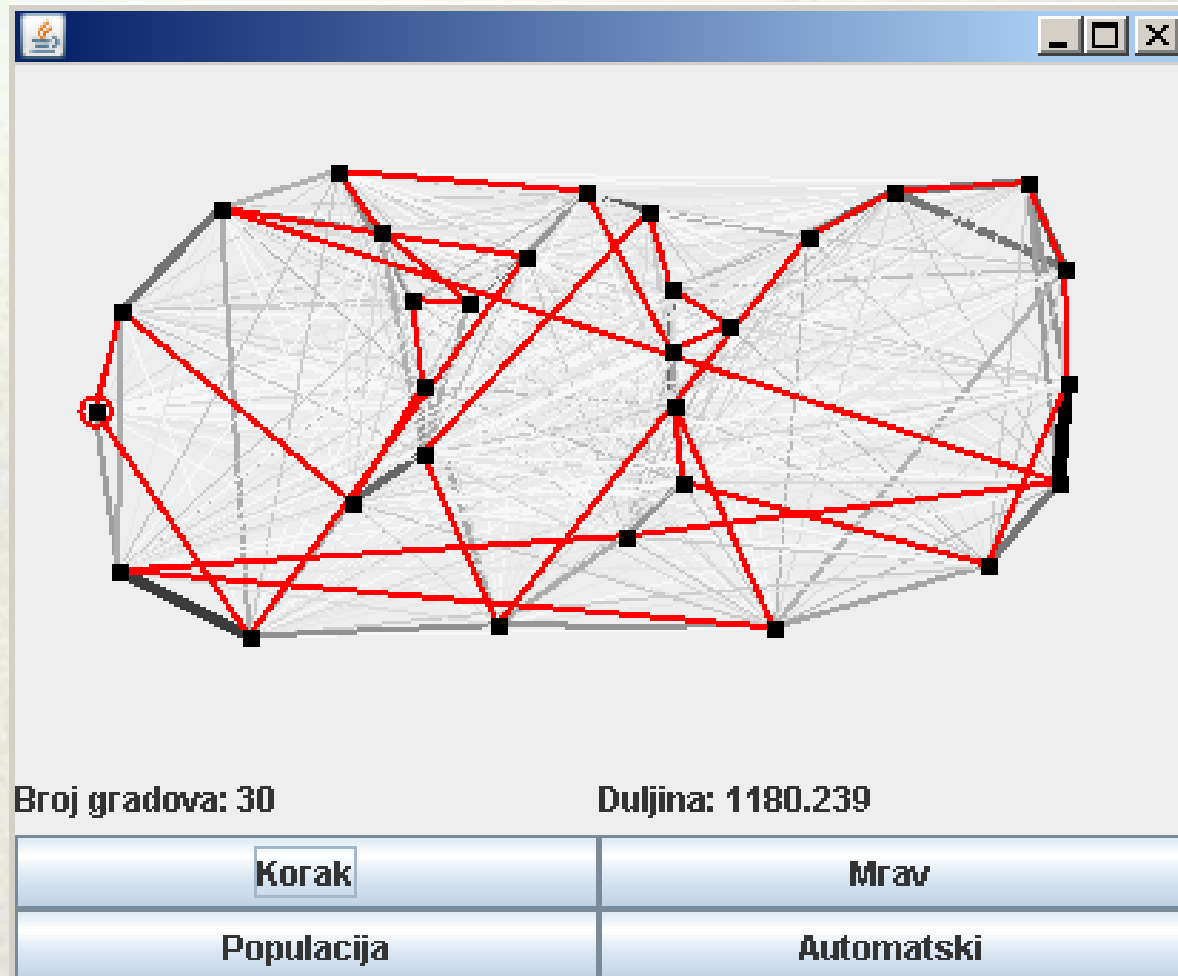
- Procedura: Update\_pheromons
  - Funkcija za odabranog mrava dodaje nove feromonske tragove iznosa:

$$\Delta\tau_{ij}^k = \begin{cases} 1/C^k, & \text{if edge } i-j \text{ is on the path of ant } k \\ 0, & \text{otherwise} \end{cases}$$

- Novo stanje je tada:

$$\tau_{ij} \leftarrow \tau_{ij} + \sum_{k=1}^m \Delta\tau_{ij}^k$$

# Ant colony optimization





# Conclusion

- Algorithms inspired by nature – a very vivid area of research!
- Methods that can efficiently tackle problems that were previously unsolvable
- New algorithms emerging (e.g. Bee Colony Optimization, Intelligent Water Drops, ...)

# Links

- Video about an ant colony  
<http://www.inquisitr.com/14238/holy-crap-billions-of-ants-in-one-colony/>



# Links

- Materials

<http://java.zemris.fer.hr/nastava/ui/>

- Implementations

[http://java.zemris.fer.hr/nastava/ui/ev  
oAlg.zip](http://java.zemris.fer.hr/nastava/ui/ev<br/>oAlg.zip)



## Ant Colony Optimization

Morça Dorigo and Thomas Stützle

