



Biologically Inspired Algorithms

Ant Colony Optimization

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History and Use

- History
- Principle
- Example

- Introduced in 1992
- Author: Marco Dorigo
- Based on the behavior of the real ant colonies
- Originally applied to Travelling Salesman Problem (TSP)
- Developed to solve discrete optimization problems

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Inspiration by Real Ants

- History
- Principle
- Example

- Each ant produces **pheromones** while travelling from the nest to food and from the place with the food to the nest → type of communication with the other ants from the colony
- The movement of the ants is random, however, during time, their decision is influenced by the pheromones
- On the shortest path, the pheromones are accumulated
- Important feature of pheromones: **vaporization**

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Principle

- History
- Principle
- Example

- Thanks to vaporization, the paths which are not the shortest are not so attractive for the ants
- In the TSP:
 - The probability that an ant k at node r will choose the destination node s is

$$p_k(r, s) = \begin{cases} \frac{\tau(r, s)^\alpha \eta(r, s)^\beta}{\sum_{u \in M_k} \tau(r, u)^\alpha \eta(r, u)^\beta}, & \text{for } s \in M_k \\ 0, & \text{otherwise} \end{cases}$$

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Principle

- History
- Principle
- Example

$$\text{pheromone} \rightarrow p_k(r, s) = \begin{cases} \frac{\tau(r, s)^\alpha \eta(r, s)^\beta}{\sum_{u \in M_k} \tau(r, u)^\alpha \eta(r, u)^\beta}, & \text{for } s \in M_k \\ 0, & \text{otherwise} \end{cases} \quad \frac{1}{d(r, s)}$$

- where:
 - α ... the degree of importance of pheromone
 - β ... degree of importance of the distance
 - $u \in M_k$... is a choice that belongs ant k (neighborhood) when he was at no r
 - Neighborhood of ant k at node r contains all the nodes that are incident with the node r except already visited nodes

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Principle of Pheromones recalculation I

- History
- Principle
- Example

- When all ants finished their journey, the pheromones must be recalculated:
 1. Vaporization

$$\tau_{r,s} = \tau_{r,s} * \rho \leftarrow \text{vaporization coefficient}$$

2. Calculation with pheromones left by all ants during their journey

$$\tau_{r,s} = \tau_{r,s} + \frac{Q}{f(s)}$$

Where :

Q ... is a constant (usually 1)

f ... the value of the objective function of the solution s

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Principle of Pheromones recalculation II

- History
- Principle
- Example

- When ant k passing the edge, it will leave a pheromone on this edge. The amount of pheromone contained in the segment ij after passed by ant k is given by:

$$\tau_{i,j} \leftarrow \tau_{i,j} + \Delta\tau^k$$

where:

$$\Delta\tau_{i,j}^k = \begin{cases} \frac{c * f_{best}}{f_{worst}} & \text{if } (ij) \in \text{best global path} \\ 0, & \text{otherwise} \end{cases}$$

where :

f_{best} ... is the best value of the objective function

c ... constant

f_{worst} ... is the worst value of the objective function

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Principle of Pheromones recalculation II

- History
- Principle
- Example

- With increasing value of pheromone on segment i, j the probability of this segment to be chosen by ants at the next iteration increases
- When a node is passed, then the pheromone evaporation will occur using the following equation:

$$\tau_{i,j} \leftarrow (1 - \rho)\tau_{i,j}, \forall (i, j) \in A$$

Where:

$\rho \in [0,1]$... evaporation rate parameter

A ... segments that have been passed by ant k as part of the path from the nest to the food

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Example

Step 1

- History
- Principle
- Example

- 5 cities

- Distance matrix for these cities is:

0	10	12	11	14
10	0	13	15	8
12	13	0	9	14
11	15	9	0	16
14	8	14	16	0

- Matrix of inverse distances
(visibility matrix):

0	0.1000	0.0833	0.0909	0.0174
0.1000	0	0.0769	0.0667	0.1250
0.0833	0.0769	0	0.1111	0.0714
0.0909	0.0667	0.1111	0	0.0625
0.0714	0.125	0.0714	0.0625	0

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Example

Step 1

- History
- Principle
- Example

- 5 cities, city 1 is a departure city. Since city 1 is chosen as the beginning , it cannot be chosen again → visibility of the city is **0**

- Initial pheromone matrix

1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1

- Visibility matrix

0	0.1000	0.0833	0.0909	0.0174
0	0	0.0769	0.0667	0.1250
0	0.0769	0	0.1111	0.0714
0	0.0667	0.1111	0	0.0625
0	0.125	0.0714	0.0625	0

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Example

Step 2

- History
- Principle
- Example

- We calculate the possibility to visit other city using the formula

$$p_k(r, s) = \begin{cases} \frac{\tau(r, s)^\alpha \eta(r, s)^\beta}{\sum_{u \in M_k} \tau(r, u)^\alpha \eta(r, u)^\beta}, & \text{for } s \in M_k \\ 0, & \text{otherwise} \end{cases}$$

$\tau(1, s)^1 \eta(1, s)^2$:

- $1 * 0.1000^2 = 0.0100$
- $1 * 0.0833^2 = 0.0069$
- $1 * 0.0909^2 = 0.0083$
- $1 * 0.0714^2 = 0.0051$

$$\sum_{s \in M_k} \tau(1, s)^1 \eta(1, s)^2 = 0.0303$$

- Visibility matrix

0	0.1000	0.0833	0.0909	0.0174
0	0	0.0769	0.0667	0.1250
0	0.0769	0	0.1111	0.0714
0	0.0667	0.1111	0	0.0625
0	0.125	0.0714	0.0625	0

[1]

Example

Step 2

- History
- Principle
- Example

$$\tau(1, s)^1 \eta(1, s)^2:$$

- $1 * 0.1000^2 = 0.0100$
- $1 * 0.0833^2 = 0.0069$
- $1 * 0.0909^2 = 0.0083$
- $1 * 0.0714^2 = 0.0051$

$$\sum_{s \in M_k} \tau(1, s)^1 \eta(1, s)^2 = 0.0303$$

The probabilities:

$$\text{City 1 to 2: } \frac{0.01}{0.0303} = 0.3299$$

$$\text{City 1 to 3: } \frac{0.0069}{0.0303} = 0.2291$$

$$\text{City 1 to 4: } \frac{0.0083}{0.0303} = 0.2727$$

$$\text{City 1 to 5: } \frac{0.0051}{0.0303} = 0.1683$$

[1]

- Visibility matrix

0	0.1000	0.0833	0.0909	0.0174
0	0	0.0769	0.0667	0.1250
0	0.0769	0	0.1111	0.0714
0	0.0667	0.1111	0	0.0625
0	0.125	0.0714	0.0625	0

Example

Step 2

- History
- Principle
- Example

- The probabilities:

$$\text{City 1 to 2: } \frac{0.01}{0.0303} = 0.3299$$

$$\text{City 1 to 3: } \frac{0.0069}{0.0303} = 0.2291$$

$$\text{City 1 to 4: } \frac{0.0083}{0.0303} = 0.2727$$

$$\text{City 1 to 5: } \frac{0.0051}{0.0303} = 0.1683$$

- The cumulative numbers of these probabilities:

- City 2: 0.3299
- City 3: 0.5590
- City 4: 0.8317
- City 5: 1

[1]

Example

Step 3

- History
- Principle
- Example

- The cumulative numbers of these probabilities:
 - City 2: 0.3299
 - City 3: 0.5590
 - City 4: 0.8317
 - City 5: 1
- Generate random number $r \in [0,1]$: suppose we have generated number $r = 0.6841$
- Compare r with cumulative numbers:
 - **$0.5590 < r < 0.8317 \Rightarrow$** an ant will visit the **city 4**

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Example

Step 4

- History
- Principle
- Example

- The city 4 was visited \Rightarrow the visibility matrix must be adjusted:

0	0.1000	0.0833	0	0.0174
0	0	0.0769	0	0.1250
0	0.0769	0	0	0.0714
0	0.0667	0.1111	0	0.0625
0	0.125	0.0714	0	0

- Now, the proces of the calculation of the probability of visiting the neighbor city will be repeated

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Example

Step 5

- History
- Principle
- Example

$$\tau(4, s)^1 \eta(4, s)^2:$$

- $1 * 0.0667^2 = 0.0044$
- $1 * 0.1111^2 = 0.0123$
- $1 * 0.07142^2 = 0.0039$

$$\sum_{s \in M_k} \tau(1, s)^1 \eta(1, s)^2 = 0.0207$$

The probabilities:

$$\text{City 4 to 2: } \frac{0.0044}{0.0207} = 0.2147$$

$$\text{City 4 to 3: } \frac{0.0123}{0.0207} = 0.2291$$

- Visibility matrix

0	0.1000	0.0833	0	0.0174
0	0	0.0769	0	0.1250
0	0.0769	0	0	0.0714
0	0.0667	0.1111	0	0.0625
0	0.125	0.0714	0	0

$$\text{City 4 to 5: } \frac{0.0039}{0.0207} = 0.1887$$

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Example

Step 5

- History
- Principle
- Example

- The cumulative numbers of these probabilities:
 - City 2: 0.2147
 - City 3: 0.8113
 - City 5: 1
- Generate random number $r \in [0,1]$: suppose we have generated number $r = 0.4024$. Compare r with cumulative numbers:
 - **$0.2147 < r < 0.8113 \Rightarrow$ an ant will visit the **city 3****
- For now, we have path $1 \rightarrow 4 \rightarrow 3$

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Example

Step 5

- History
- Principle
- Example

- This process is repeated until all ants have their own paths
- Remember: Each ant starts from another city
- Suppose that we have the paths as follows:
 - Ant 1: 1 → 4 → 3 → 5 → 2 → 1 Total distance: 52
 - Ant 2: 1 → 4 → 2 → 5 → 3 → 1 Total distance: 60
 - Ant 3: 1 → 4 → 5 → 2 → 3 → 1 Total distance: 60
- Use total distances (objective function evaluation) to recalculate the pheromones on the edges:

$$\tau_{r,s} \leftarrow (1 - \rho)\tau_{r,s} + \sum_{k=1}^N \Delta\tau_{r,s}^k \quad [1]$$

Example

Step 6

- History
- Principle
- Example

- Use total distances (objective function evaluation) to recalculate the pheromones on the edges:

$$\tau_{r,s} \leftarrow (1 - \rho)\tau_{r,s} + \sum_{k=1}^N \Delta\tau_{r,s}^k$$

ρ ... evaporation coefficient equals to 0.5

- Pheromone matrix recalculated based on the Ant 1:

0.5000	0.5000	0.5000	0.5192	0.5000
0.5192	0.5000	0.5000	0.5000	0.5000
0.5000	0.5000	0.5000	0.5000	0.5192
0.5000	0.5000	0.5192	0.5000	0.5000
0.5000	0.5192	0.5000	0.5000	0.5000

$$\Delta\tau_{r,s}^k = \frac{Q}{f(s)} = \frac{1}{52}$$

[1]

Example

Step 6

- History
- Principle
- Example

- Use total distances (objective function evaluation) to recalculate the pheromones on the edges:

$$\tau_{r,s} \leftarrow (1 - \rho)\tau_{r,s} + \sum_{k=1}^N \Delta\tau_{r,s}^k$$

ρ ... evaporation coefficient equals to 0.5

- Pheromone matrix recalculated based on the Ant 2:

0.5000	0.5000	0.5000	0.5359	0.5000
0.5192	0.5000	0.5000	0.5000	0.5167
0.5167	0.5000	0.5000	0.5000	0.5192
0.5000	0.5167	0.5192	0.5000	0.5000
0.5000	0.5192	0.5167	0.5000	0.5000

$$\Delta\tau_{r,s}^k = \frac{Q}{f(s)} = \frac{1}{60}$$

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Example

Step 6

- History
- Principle
- Example

- Use total distances (objective function evaluation) to recalculate the pheromones on the edges:

$$\tau_{r,s} \leftarrow (1 - \rho)\tau_{r,s} + \sum_{k=1}^N \Delta\tau_{r,s}^k$$

ρ ... evaporation coefficient equals to 0.5

- Pheromone matrix recalculated based on the Ant 3:

0.5000	0.5000	0.5000	0.5526	0.5000
0.5192	0.5000	0.5167	0.5000	0.5167
0.5334	0.5000	0.5000	0.5000	0.5192
0.5000	0.5167	0.5192	0.5000	0.5167
0.5000	0.5359	0.5167	0.5000	0.5000

$$\Delta\tau_{r,s}^k = \frac{Q}{f(s)} = \frac{1}{60}$$

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Example

Step 7

- History
- Principle
- Example

- Repeat the proces until the number of iterations equals to the maximum number of iterations

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Literature

[1] Budi Santosa: Tutorial on Ant Colony Optimization, Institut Teknologi Sepuluh Nopember, ITS, bsantosa.com