

# Ant Colony Algorithms for Combinatorial Optimization

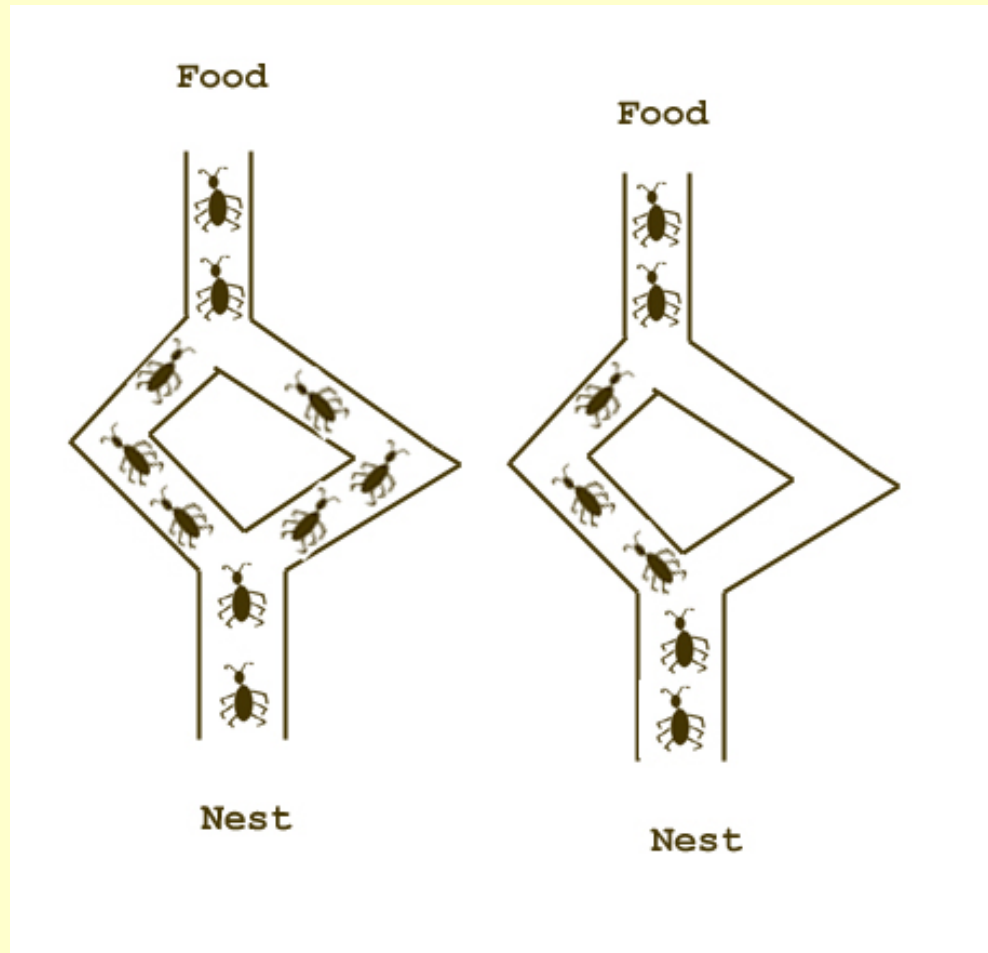
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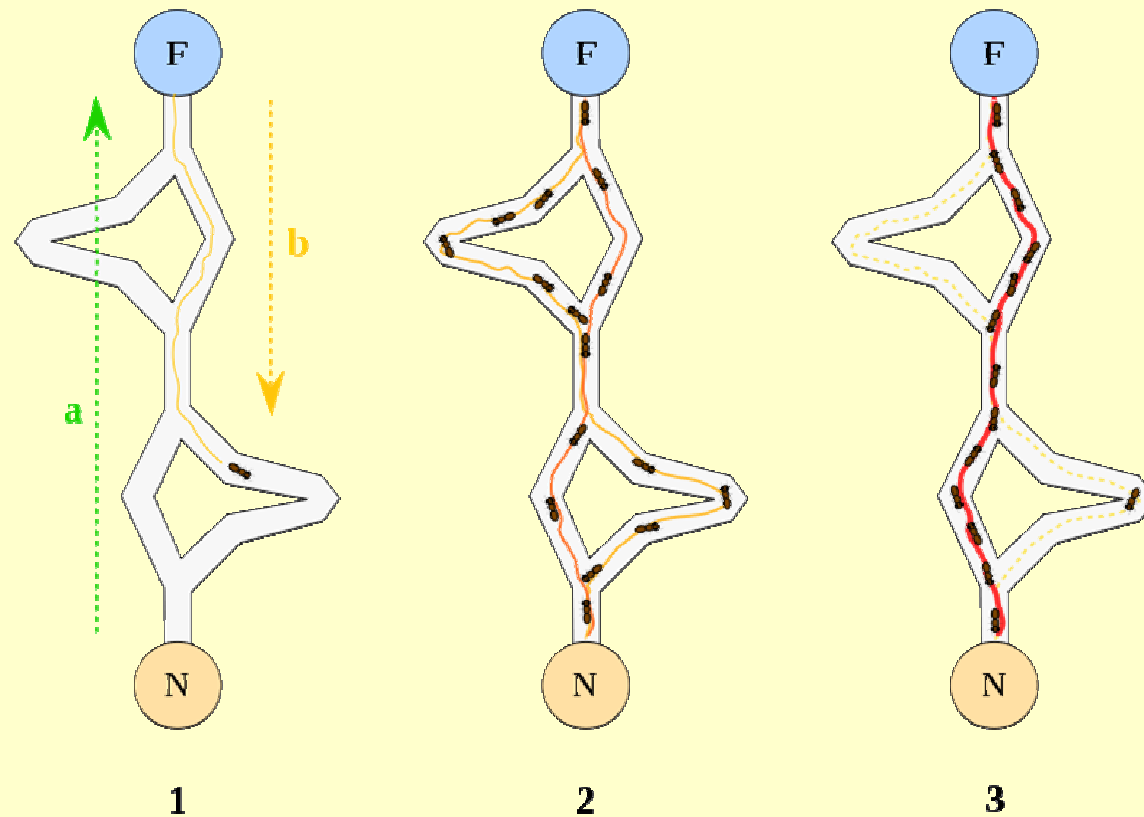
# Metaheuristics

- A **metaheuristics** are methods for solving a very general class of computational problems by combining user-given black-box procedures in the hope of obtaining a more efficient or more robust procedure. The name combines the Greek prefix "meta" ("beyond", here in the sense of "higher level") and "heuristic" (from εὕρισκειν, *heuriskein*, "to find").
- Metaheuristics are generally applied to problems for which there is no satisfactory problem-specific algorithm or heuristic; or when it is not practical to implement such a method. Most commonly used metaheuristics are targeted to combinatorial optimization problems, but of course can handle any problem that can be recast in that form, such as solving boolean equations

# Ant Colony Optimization



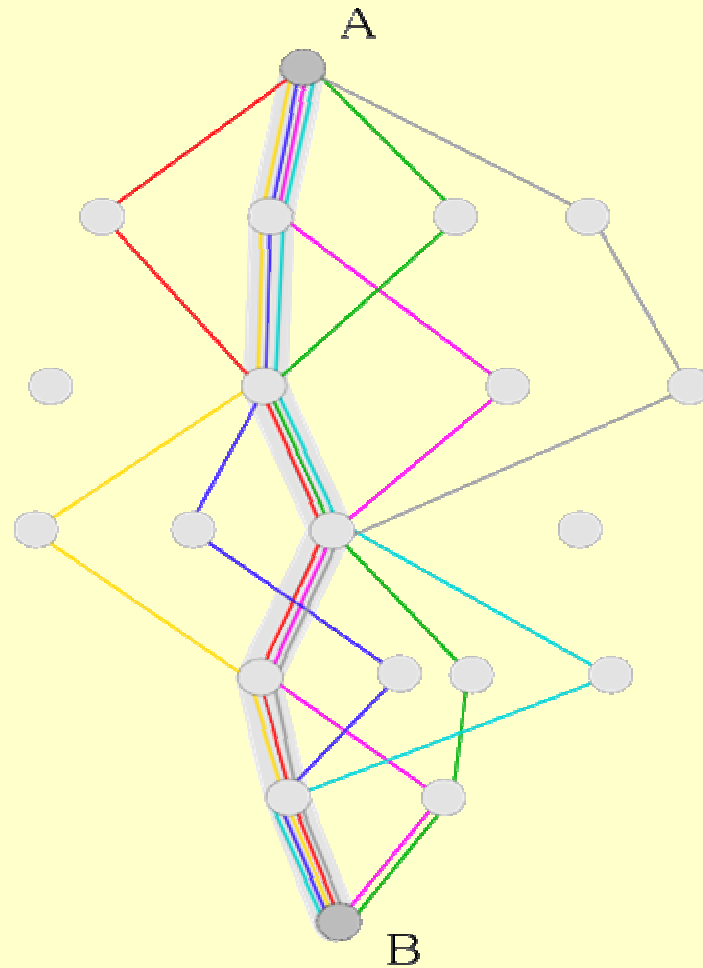
# Ant Colony Optimization



# Hensel and Grethen



# Graph of the Problem



# Ant Colony Optimization

## Procedure ACO

### Begin

initialize the pheromone

**while** stopping criterion not satisfied **do**  
position each ant on a starting node

**repeat**

**for** each ant **do**

chose next node

**end for**

**until** every ant has build a solution

update the pheromone

**end while**

**end**



# Transition Probability



$$\text{Pr ob}_{ij}^k(t) = \begin{cases} \frac{\tau_{ij} \eta_{ij}}{\sum_{b \in \text{allowed}_k(t)} \tau_{ib} \eta_{ib}} & \text{if } j \in \text{allowed}_k(t) \\ 0 & \text{otherwise} \end{cases}$$



# Pheromone Updating



$$\tau_{ij} \leftarrow \rho \tau_{ij} + \Delta \tau_{ij}$$

$0 < \rho < 1$  – evaporation

# Ant Colony Algorithms

- Ant System
- Ant Colony System
- ACO with elitist ants
- Max-Min Ant System
- ACO with additional reinforcement



# Ant System



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

$$\Delta \tau_{ij} = \begin{cases} 1 / F & \text{if minimisation problem} \\ F & \text{if maximization problem} \end{cases}$$

$$0 < \rho < 1$$

# Ant Colony System

$$\tau_{ij} \leftarrow (1 - \rho)\tau_{ij} + \rho\tau_0$$

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

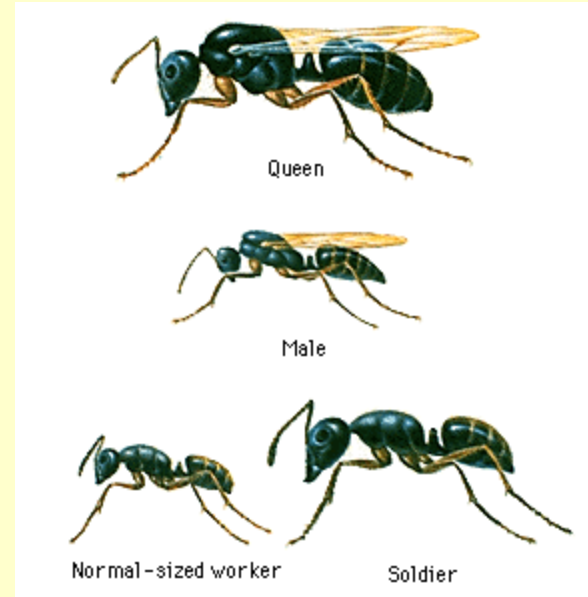
$$\Delta\tau_{ij} = \begin{cases} L_{gb} & \text{if } (i, j) \in \text{global best} \\ 0 & \text{otherwise} \end{cases}$$



# ACO with Elitist Ants

$$\tau_{ij} \leftarrow \rho \tau_{ij}$$

$$\tau_{ij} \leftarrow \tau_{ij} + (1 - \rho)L \quad \text{if the ant is elit}$$



# Max-Min Ant System



$\tau_{\max}$        $\tau_{\min}$

*If  $\tau_{ij} < \tau_{\min}$  then  $\tau_{ij} = \tau_{\min}$*

*If  $\tau_{ij} > \tau_{\max}$  then  $\tau_{ij} = \tau_{\max}$*

# ACO with Additional Reinforcement

$$\tau_{ij} \leftarrow \rho \tau_{ij} + \Delta \tau_{ij}$$

$$0 < \rho < 1 \quad - \text{evaporation}$$

$$\tau_{ij} \leftarrow \alpha \tau_{ij} + q \tau_{\max}$$

$$\alpha = \begin{cases} 1 - \text{if unused movements are evaporated} \\ \rho - \text{otherwise} \end{cases}$$

$$q \leq \rho$$

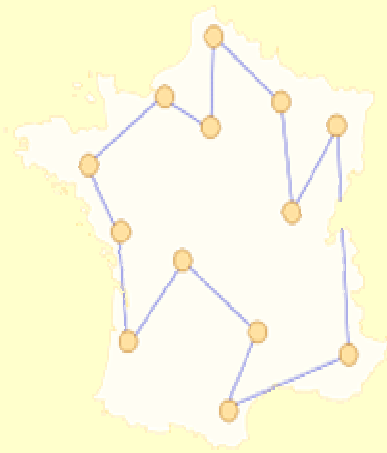


# Traveling Salesman Problem

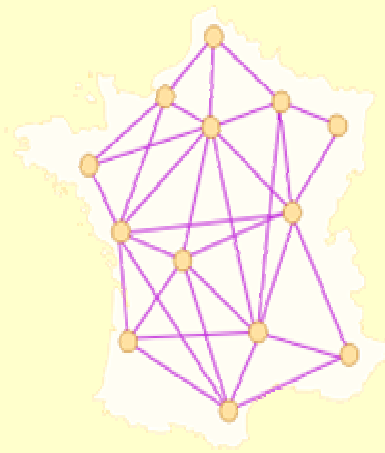
- It must visit each city exactly once;
- A distant city has less chance of being chosen (the visibility);
- The more intense the pheromone trail laid out on an edge between two cities, the greater the probability that that edge will be chosen;
- Having completed its journey, the ant deposits more pheromones on all edges it traversed, if the journey is short;
- After each iteration, trails of pheromones evaporate.



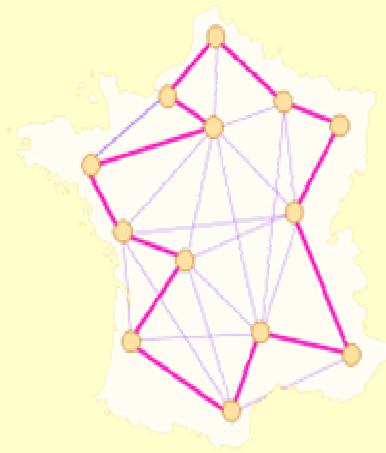
# Traveling Salesman Problem



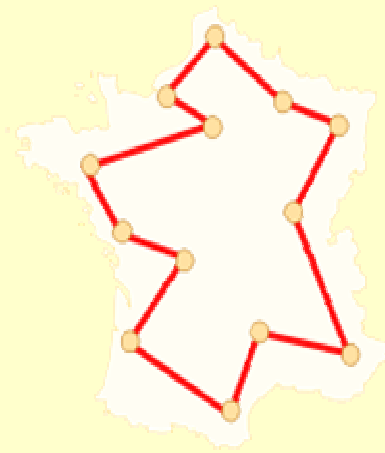
**1**



**2**



**3**



**4**

# Subset of Nodes

- $N1 \rightarrow S1$
- $N2 \rightarrow S2 \dots$
- $Nn \rightarrow Sn$
- $N(n+1) \rightarrow S1 \dots$



# Subset Estimation



$$D_j(i) = \varphi.D_j(i) + (1 - \varphi).F_j(i)$$

$$i \geq 1, \quad 1 \leq j \leq n,$$

$$F_j(i) = \begin{cases} \frac{f_{jA}}{k_j} & \text{if } k_j \neq 0 \\ F_j(i-1) & \text{otherwise} \end{cases}$$

$$E_j(i) = \varphi.E_j(i) + (1 - \varphi).G_j(i)$$

$$G_j(i) = \begin{cases} \frac{g_{jB}}{k_j} & \text{if } k_j \neq 0 \\ G_j(i-1) & \text{otherwise} \end{cases}$$

# Start Strategies



1. If  $E_j(i)/D_j(i) > E$  then the subset  $j$  is forbidden for current iteration;
2. If  $E_j(i)/D_j(i) > E$  then the subset  $j$  is forbidden for current simulation
3. If  $E_j(i)/D_j(i) > E$  then the subset  $j$  is forbidden for  $K1$  iterations
4. Let  $r1 \in (0.5, 1)$  and  $r2 \in (0, 1)$  are random numbers. If  $r2 > r1$  randomly choose a node from  $\{j | D_j(i) > D\}$
5. The same as in case 4. but  $D$  increase with  $r3$  every iteration

# Thank for Your Attention

