

# Ant colony Optimization Algorithms : Introduction and Beyond

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Artificial Intelligence Seminar 2009

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- Meta-heuristic Optimization
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## Main ACO Algorithms

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- Ant Colony System
- MAX-MIN Ant System

3

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- Advantages
- Disadvantages

# What is Ant Colony Optimization?

## ACO

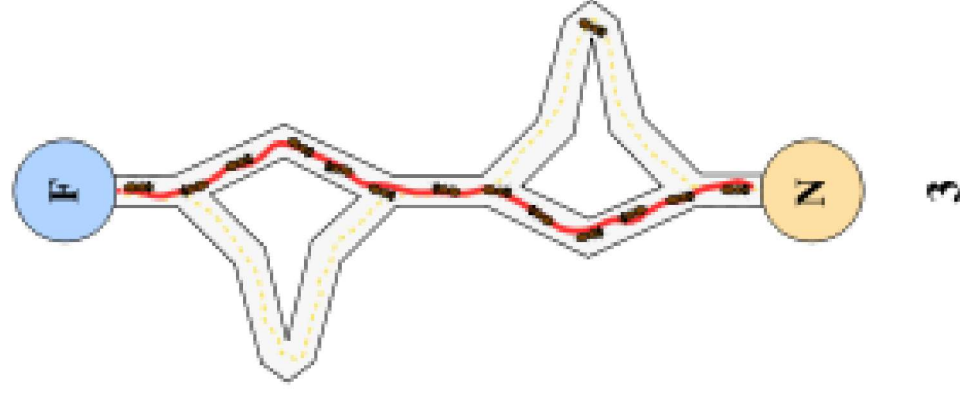
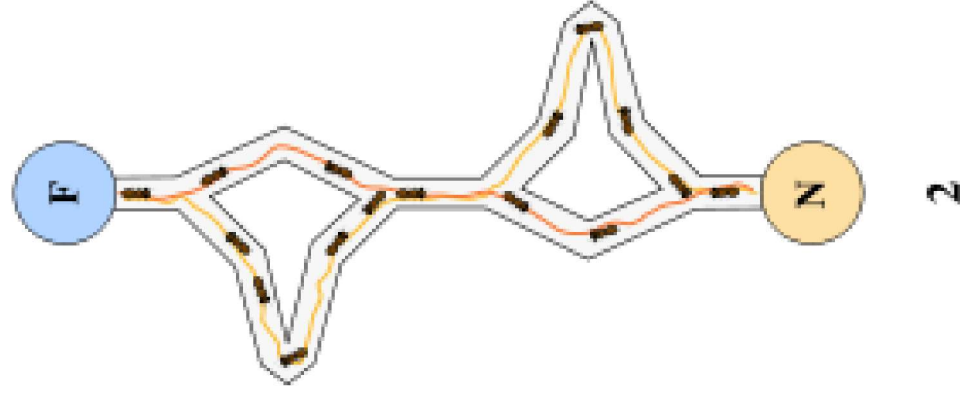
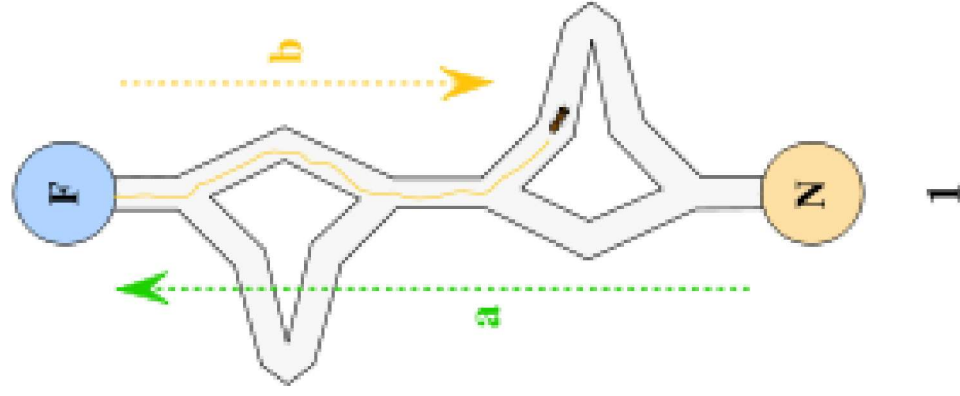
- **Probabilistic** technique.
- Searching for **optimal path** in the graph based on behaviour of **ants** seeking a path between their colony and source of food.
- **Meta-heuristic** optimization

# ACO Concept

## Overview of the Concept

- Ants navigate from nest to food source. **Ants are blind!**
- Shortest path is discovered via **pheromone** trails.
- Each ant moves at random
- Pheromone is deposited on path
- More pheromone on path increases probability of path being followed

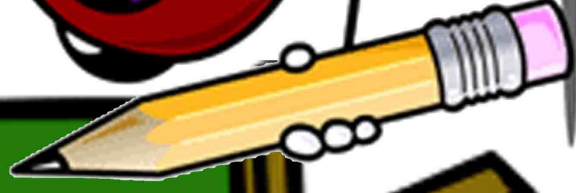
## Ant Colony Optimization



## Ant Colony Optimization

This problem might seem trivial to us, but to certain inferior species, this is considered “hard”

## Traveling Salesman Problem



Art by Pratik

# ACO System

## Overview of the System

- Virtual trail accumulated on path segments
- Path selected at random based on amount of "trail" present on possible paths from starting node

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## Overview of the System

- Virtual trail accumulated on path segments
- Path selected at random based on amount of "trail" present on possible paths from starting node
- Ant reaches next node, selects next path
- Continues until reaches starting node
- Finished tour is a solution.
- Tour is analyzed for optimality

# Meta-heuristic

- 1 Heuristic method for solving a very general class of computational problems by combining user-given heuristics in the hope of obtaining a more efficient procedure.

# Meta-heuristic

- 1 Heuristic method for solving a very general class of computational problems by combining user-given heuristics in the hope of obtaining a more efficient procedure.
- 2 ACO is meta-heuristic
- 3 Soft computing technique for solving hard discrete optimization problems

# History



- 1 **Ant System** was developed by **Marco Dorigo** (Italy) in his PhD thesis in 1992.
- 2 **Max-Min Ant System** developed by Hoos and Stützle in 1996
- 3 **Ant Colony** was developed by Gambardella Dorigo in 1997

# The ACO Meta-heuristic

## ACO

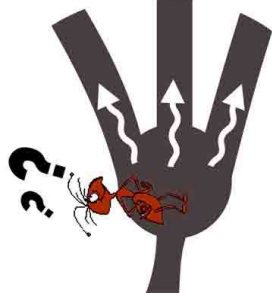
Set Parameters, Initialize pheromone trails

## SCHEDULE ACTIVITIES

- 1 Construct Ant Solutions
- 2 Daemon Actions (optional)
- 3 Update Pheromones

Virtual trail accumulated on path segments

# ACO - Construct Ant Solutions



## ACO - Construct Ant Solutions

An ant will move from node  $i$  to node  $j$  with probability

$$p_{i,j} = \frac{(\tau_{i,j}^{\alpha})(\eta_{i,j}^{\beta})}{\sum (\tau_{i,j}^{\alpha})(\eta_{i,j}^{\beta})}$$

where

$\tau_{i,j}$  is the amount of pheromone on edge  $i,j$

$\alpha$  is a parameter to control the influence of  $\tau_{i,j}$

$\eta_{i,j}$  is the desirability of edge  $i,j$  (typically  $1/d_{i,j}$ )

$\beta$  is a parameter to control the influence of  $\eta_{i,j}$

## ACO - Pheromone Update

### ACO - Pheromone Update

Amount of pheromone is updated according to the equation

$$\tau_{i,j} = (1 - \rho)\tau_{i,j} + \Delta\tau_{i,j}$$

where

$\tau_{i,j}$  is the amount of pheromone on a given edge  $i, j$

$\rho$  is the rate of pheromone evaporation

$\Delta\tau_{i,j}$  is the amount of pheromone deposited, **typically** given by

$$\Delta\tau_{i,j}^k = \begin{cases} 1/L_k & \text{if ant } k \text{ travels on edge } i, j \\ 0 & \text{otherwise} \end{cases}$$

where  $L_k$  is the cost of the  $k^{th}$  ant's tour (typically length).

# ACO

## ACO

- Many special cases of the ACO metaheuristic have been proposed.
- The three most successful ones are: Ant System, Ant Colony System (ACS), and MAX-MIN Ant System (MMAS).
- For illustration, example problem used is Travelling Salesman Problem.



# ACO - Ant System

## ACO - Ant System

- First ACO algorithm to be proposed (1992)
- Pheromone values are updated by all the ants that have completed the tour.

$$\tau_{ij} \leftarrow (1 - \rho) \cdot \tau_{ij} + \sum_{k=1}^m \Delta\tau_{ij}^k,$$

where

$\rho$  is the evaporation rate

$m$  is the number of ants

$\Delta\tau_{ij}^k$  is pheromone quantity laid on edge  $(i, j)$  by the  $k^{th}$  ant

$$\Delta\tau_{i,j}^k = \begin{cases} 1/L_k & \text{if ant } k \text{ travels on edge } i, j \\ 0 & \text{otherwise} \end{cases}$$

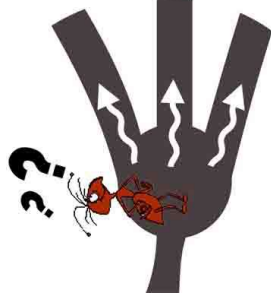
where  $L_k$  is the tour length of the  $k^{th}$  ant.

# ACO - Ant Colony System

## ACO - Ant Colony System

- First major improvement over Ant System
- Differences with Ant System:
  - 1 Decision Rule - Pseudorandom proportional rule
  - 2 Local Pheromone Update
  - 3 Best only offline Pheromone Update

# ACO - Ant Colony System



## ACO - Ant Colony System

- Ants in ACS use the **pseudorandom proportional rule**
- Probability for an ant to move from city  $i$  to city  $j$  depends on a random variable  $q$  uniformly distributed over  $[0, 1]$ , and a parameter  $q_0$ .
- If  $q \leq q_0$ , then, among the feasible components, the component that maximizes the product  $\tau_{ij}\eta_{ij}^\beta$  is chosen, otherwise the same equation as in Ant System is used.
- This rule favours exploitation of pheromone information

# ACO - Ant Colony System

## ACO - Ant Colony System

- Diversifying component against exploitation: **local pheromone update**.
- The local pheromone update is performed by all ants after each step.
- Each ant applies it only to the last edge traversed:

$$\tau_{ij} = (1 - \varphi) \cdot \tau_{ij} + \varphi \cdot \tau_0$$

where

$\varphi \in (0, 1]$  is the pheromone decay coefficient

$\tau_0$  is the initial value of the pheromone (value kept small

*Why?*)

## ACO - Ant Colony System

### ACO - Ant Colony System

- Best only offline pheromone update after construction
- Offline pheromone update equation

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

where

$$\tau_{ij}^{best} = \begin{cases} 1/L_{best} & \text{if best ant } k \text{ travels on edge } i, j \\ 0 & \text{otherwise} \end{cases}$$

- $L_{best}$  can be set to the length of the best tour found in the current iteration or the best solution found since the start of the algorithm.

# ACO - MAX-MIN Ant System

## ACO - MAX-MIN Ant System

- Differences with Ant System:
  - 1 Best only offline Pheromone Update
  - 2 Min and Max values of the pheromone are explicitly limited
    - $\tau_{ij}$  is constrained between  $\tau_{min}$  and  $\tau_{max}$  (explicitly set by algorithm designer).
    - After pheromone update,  $\tau_{ij}$  is set to  $\tau_{max}$  if  $\tau_{ij} > \tau_{max}$  and to  $\tau_{min}$  if  $\tau_{ij} < \tau_{min}$