

# Practical Optimization Algorithm Design

## Lesson 13: Swarm Intelligence

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## Introduction

- Swarms of tiny, simple creatures able to colossal achievements
- Emergence
- Self-Organization
- Swarm Intelligence



# Particle Swarm Optimization

- Copy the behavior with which swarms / schools / flocks in nature find food
- Search space is subset of real vectors numbers:  $\mathbb{G} \subseteq \mathbb{R}^n$
- Population = swarm of particles which move in  $\mathbb{G}$
- Genotype = position of particle  $p$
- Endogenous information = velocity vectors of particles  $p.v$   
best found position so far  $best(p).g$
- $N(p)$ : neighbors of particle  $p$

## Particle Swarm Optimization

- In each iteration for each particle  $p$ :

### 1) Velocity Update:

$$p.v = \text{randomUni}[0, c_i) * (\text{best}(p).g_i - p.g_i) + \text{randomUni}[0, d_i) * (\text{best}(N(p)).g_i - p.g_i)$$

Social  
Component

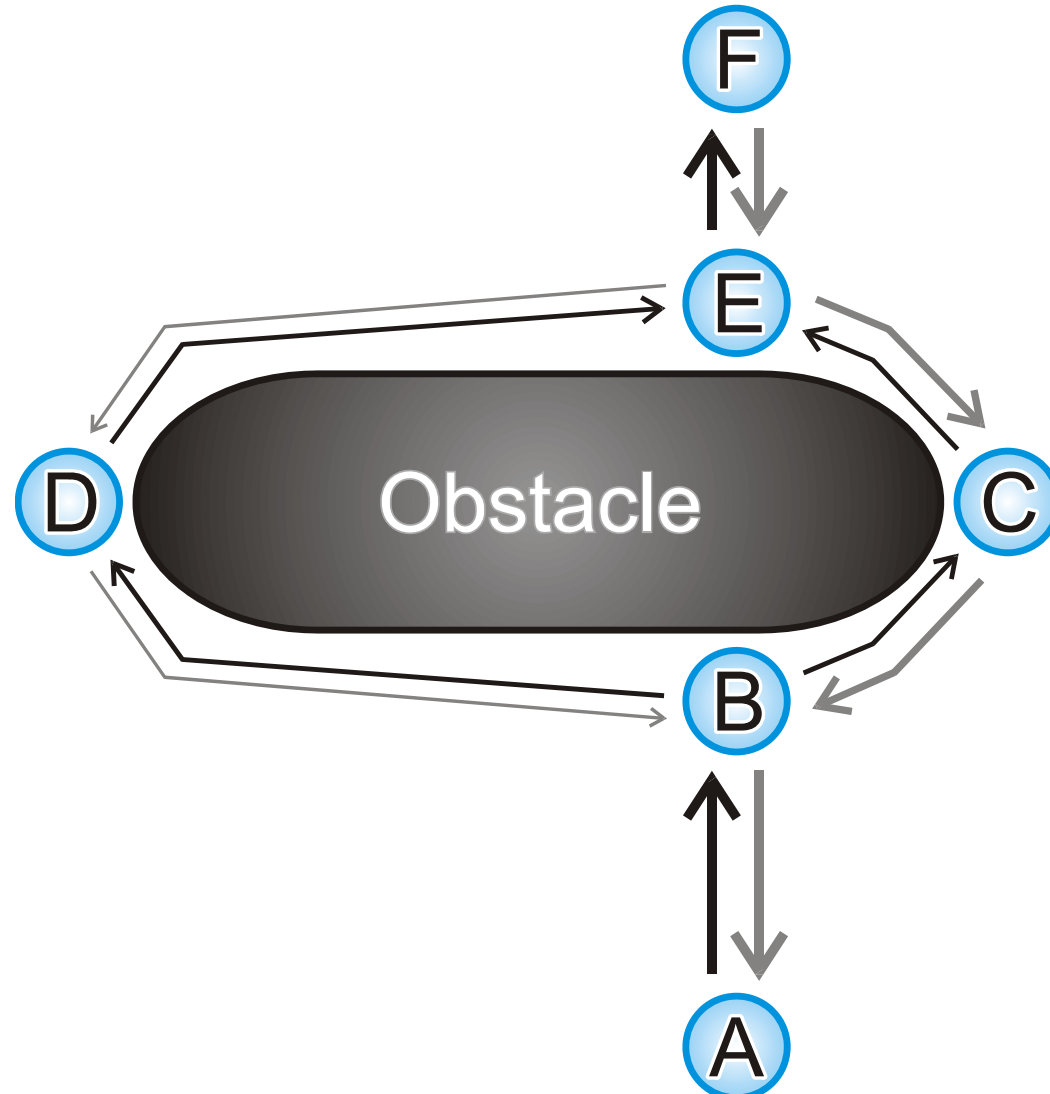
### 2) Afterwards – Position Update:

$$p.g = p.g + p.v$$

## Ant Colony Optimization

- Simulation of the way ants form a path in order to solve optimization problems which can be represented as graphs
- Ants move from one location to another
- And lay down pheromone: stigmergy = communication by modifying the environment
- Paths with more pheromone on them are more likely to be followed
- These are often the shortest paths
- Many combinatorial problems can be considered as finding the shortest path on a graph. Example: TSP

# Ant Colony Optimization



## Ant Colony Optimization

- An ant located in node  $i$  in ACO chooses the next node  $j$  where it will go according to
  - 1) the distance between  $i$  and
  - 2) the amount of pheromone on the edge connecting  $i$  and  $j$

- $$p_{i,j} = \frac{(\tau_{i,j})^\alpha (\eta_{i,j})^\beta}{\sum_{\forall k} (\tau_{i,k})^\alpha (\eta_{i,k})^\beta}$$

$p_{i,j}$  probability of an ant to go to  $j$   
(if at location  $i$ )

$\alpha, \beta$  weights

$\tau_{i,j}$  pheromone on the edge  
connecting  $i$  and  $j$

$\eta_{i,j}$  visibility of node  $j$  from node  $i$ :  
inversely proportional to the  
distance between  $i$  and  $j$

# Ant Colony Optimization

- At the end of each algorithm round, “pheromone” is dispersed and the trails are updated ( $\eta_{i,j}$  stays constant)

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

- $\rho$  is the evaporation coefficient
- $\Delta\tau_{i,j}$  is the new pheromone dispersed
- the amount usually depends on the quality of the paths the edge  $(i,j)$  was part of
- In TSP: inversely proportional to length of tour

## Summary

- Collectives of simple creatures able to colossal achievements
- Swarm Intelligence
- PSO: Copy flocking behavior
- ACO: copy ground-based movements / stigmergy

谢谢你们！

再见！