

SWARM INTELLIGENCE

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CSCI 446: Artificial Intelligence

OVERVIEW

- What is Swarm Intelligence
- Swarms in Nature
 - Ants
 - Birds
- Ant Colony Optimization
- Particle Swarm Optimization



WHAT IS SWARM INTELLIGENCE?

Swarm intelligence is how individuals, knowingly or not, cooperate together to achieve a goal.

ROOTS IN MODELS OF SOCIAL INSECT BEHAVIOR

- Ants searching for food.
- Birds flocking together.
- Termites building nest.
- Bacteria foraging for food.

SWARM INTELLIGENCE DEFINED

- Useful behavior that emerges from the cooperative efforts of a group of individual agents;
- ... in which the individual agents are largely homogeneous;
- ... in which the individual agents act asynchronously in parallel;

SWARM INTELLIGENCE DEFINED, CONT.

- in which there is little or no centralized control;
- .. in which communication between agents is largely effected by some form of stigmergy;
- ... in which there 'useful' behavior is relatively simple (finding a good place for food, or building a nest – not writing a symphony, or surviving for many years in a dynamic environment).

WHAT DO THEY HAVE IN COMMON?

- All move in groups to achieve a goal
 - Behavior of groups is special to the group
- Individuals in group act together in unison
 - Byproduct of local control of individuals
- No global control of group.

ANTS IN NATURE

- Ants individually are not so clever
 - Colony of ants can be
- Ants excel at finding the shortest and safest path to food
- Ants searching for food inspired algorithm
 - Ant Colony Optimization (ACO)

ANTS IN NATURE

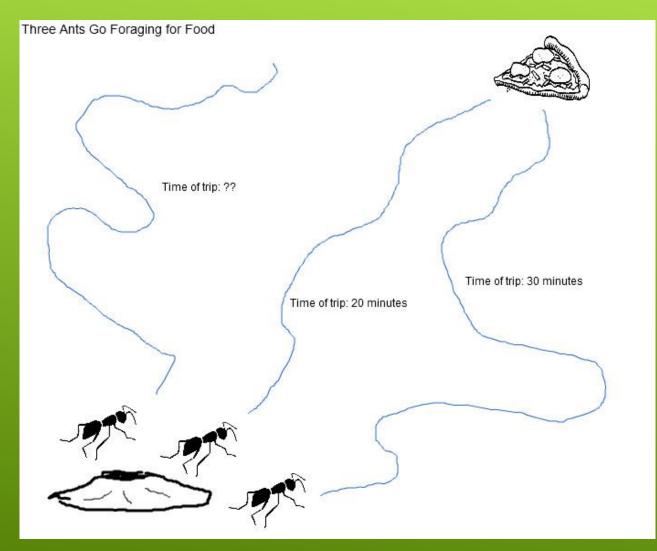
- Ants are naturally stochastic
- Have no direct forms of communication
- Communicate via
 - Touch
 - Sound
 - Pheromones
- Type of communication: Stigmergy

STIGMERGY

- An agent's actions leave sigs in the environment.
 These signs are later sensed by other agents, which in turn determine and incite their subsequent actions.
- Greek words "stigma-ergon"
 - Meaning "mark-action"

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HOW ANTS GET IT DONE

- Ants leave pheromone trails
- Strength of trail influences other ants to take path
- Ants are still stochastic by nature

BIRDS IN NATURE

- Birds flock together
 - Protection
 - Search for Food
 - Migration
- Prime example
 - Starling Murmuration



STALING MURMURATION

 <u>https://www.youtube.com/</u> <u>watch?v=eakKfY5aHmY</u>

REYNOLDS RULES: RULES OF A FLOCK

- **Cohesion**: steer towards the mean position of others, thus staying close to other flock mates
- Alignment: steer towards the mean heading of others and match velocity
- Separation: steer to avoid coming to close to others and avoid collisions

ANT COLONY OPTIMIZATION (ACO)

- Like ants in nature excel at finding shortest path
- Excellent for problems like Traveling Salesman

$$p_{ij}^{k} = \begin{cases} \frac{[\tau_{ij}]^{\alpha} [\eta_{ij}]^{\beta}}{\sum_{k \in \text{feasible}_{k}} [\tau_{ik}]^{\alpha} [\eta_{ik}]^{\beta}} & \text{if } j \in \text{feasible}_{k} \\ 0 & \text{otherwise} \end{cases}$$

- $[t_{ij}]$ reprents the pheromone trail from i to j.
- [h_{ii}] represents the heuristic value from i to j.
- a and β are influence weights of pheromones and heuristic

$$\Delta \tau_{ij}^{k}(t) = \begin{cases} 1/L^{k}(t) & \text{, if } arc(i,j) \in T^{k}(t) \\ 0 & \text{, otherwise} \end{cases}$$

- L weight of the edge between nodes i and j.
- t current iteration

PARTICLE SWARM OPTIMIZATION

- Follow Reynolds rules
 - Cohesion
 - Alignment
 - Separation
- Add new rule
 - Attraction to a target
 - Fitness function to determine how good a place is to be

PARTICLE SWARM DYNAMICS

∆v(t)

F (x(t-1), $\Delta v(t-1)$, p_{b} , p_{g})

- Particle acceleration can be a function of F
- Particle position x(t)
- Pb particle's best position
- Pg particle's neighborhood's best position

PARTICLE SWARM VELOCITY UPDATE

 $v(t) = v(t-1) + \Delta v(t-1)$

Velocity at time t is velocity at time t-1 plus the acceleration value

PARTICLE SWARM MAX VELOCITY

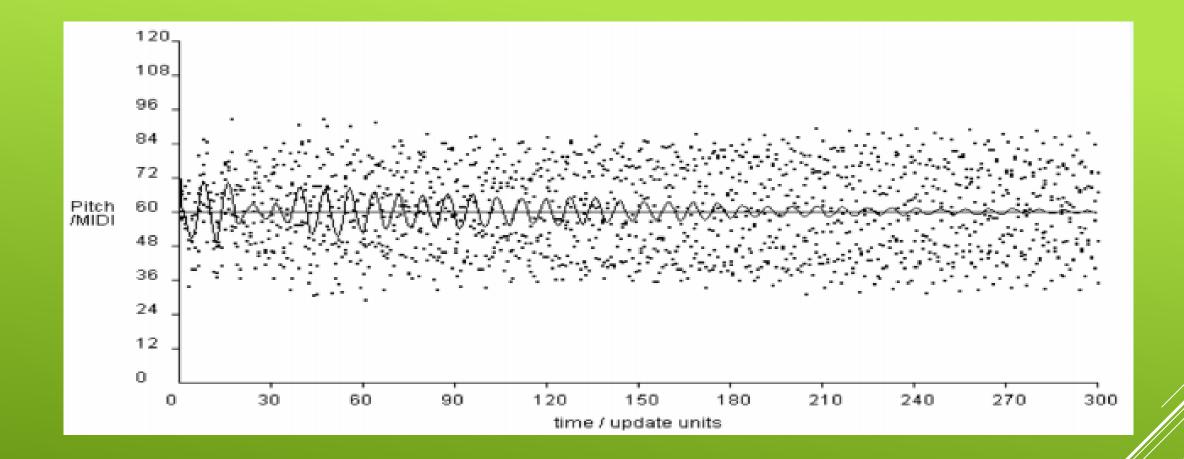
 $v(t + \varepsilon) = v(t) + \theta (|(v(t) / v_{max})| - 1) (v_{max} - v(t))$

- Provides nonlinear damping force
- Applied instantaneously
- Effect of limiting the velocity

PARTICLE SWARM POSITION UPDATE

x(t) = x(t-1) + v(t)

Particle position at time t is position at time t-1 plus velocity values



Center of swarm over time is plotted, and shows how the particles oscillate around the target

Eventually converge

CONCLUSION

- Nature inspired algorithms
- Advancement in robotics
- Used in Entertainment industry
 - Batman Returns, Lion King, Lord of the Rings
- Very good at solving specific problems that fit swarm behavior