

Grupo de Inteligencia Computacional UPV/EHU

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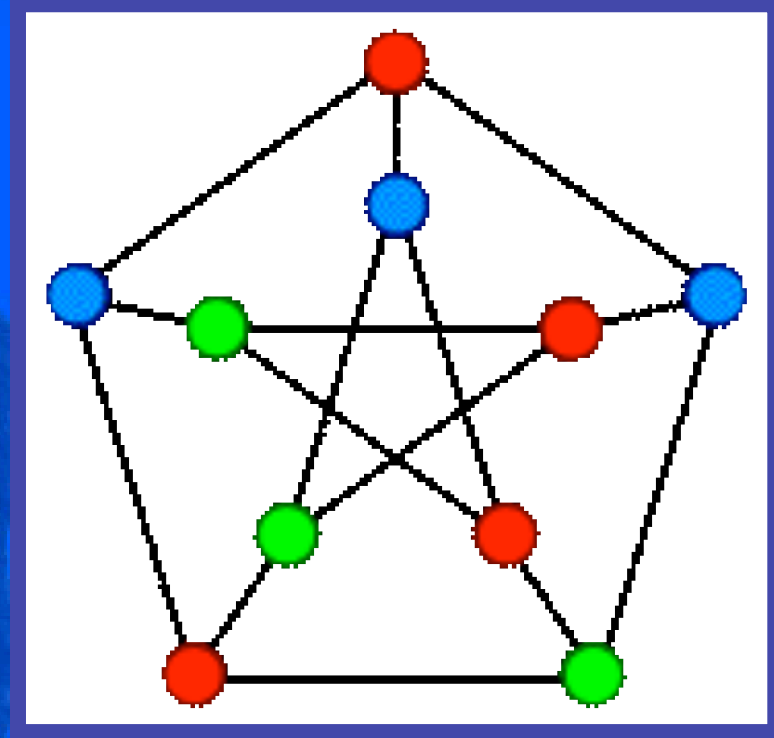


On the ability of Swarms to compute the 3-coloring of graphs

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Graph coloring problem

- Application of swarm intelligence to the k-coloring of graphs.
- Minimize the number of neighboring vertices with the same color.
- NP complete for $k \geq 3$.
- Important problems in graph theory.



Approaches to graph coloring

- Theoretical computer science: **P=NP?**
- Artificial intelligence: **heuristic search.**
- Computational Intelligence and Artificial Life:
 - Searches for non deterministic algorithms that solve the problem efficiently in polynomial times.
 - Complex systems: Neural networks, swarm intelligence, cellular automata...
 - Biological inspiration increments expressiveness of programming languages.
 - Orientation to graph drawing techniques

Self-Organizing Particle Systems (SOPS)

- Computational models of the navigation of swarms.
- As a behavior emerging from locally controlled movements.
- Based on decisions taken on local information.
- Approaches to swarm intelligence:
 - Steering behaviors: Reynolds .
 - Stirmergy: Ant colony optimization solves the travel salesman problem as an emergent behavior.
 - Optimization of abstract functions:
 - Self-organization: Particle swarm optimization (PSO).
 - Self-organisation+Evolution=Stochastic diffusion search

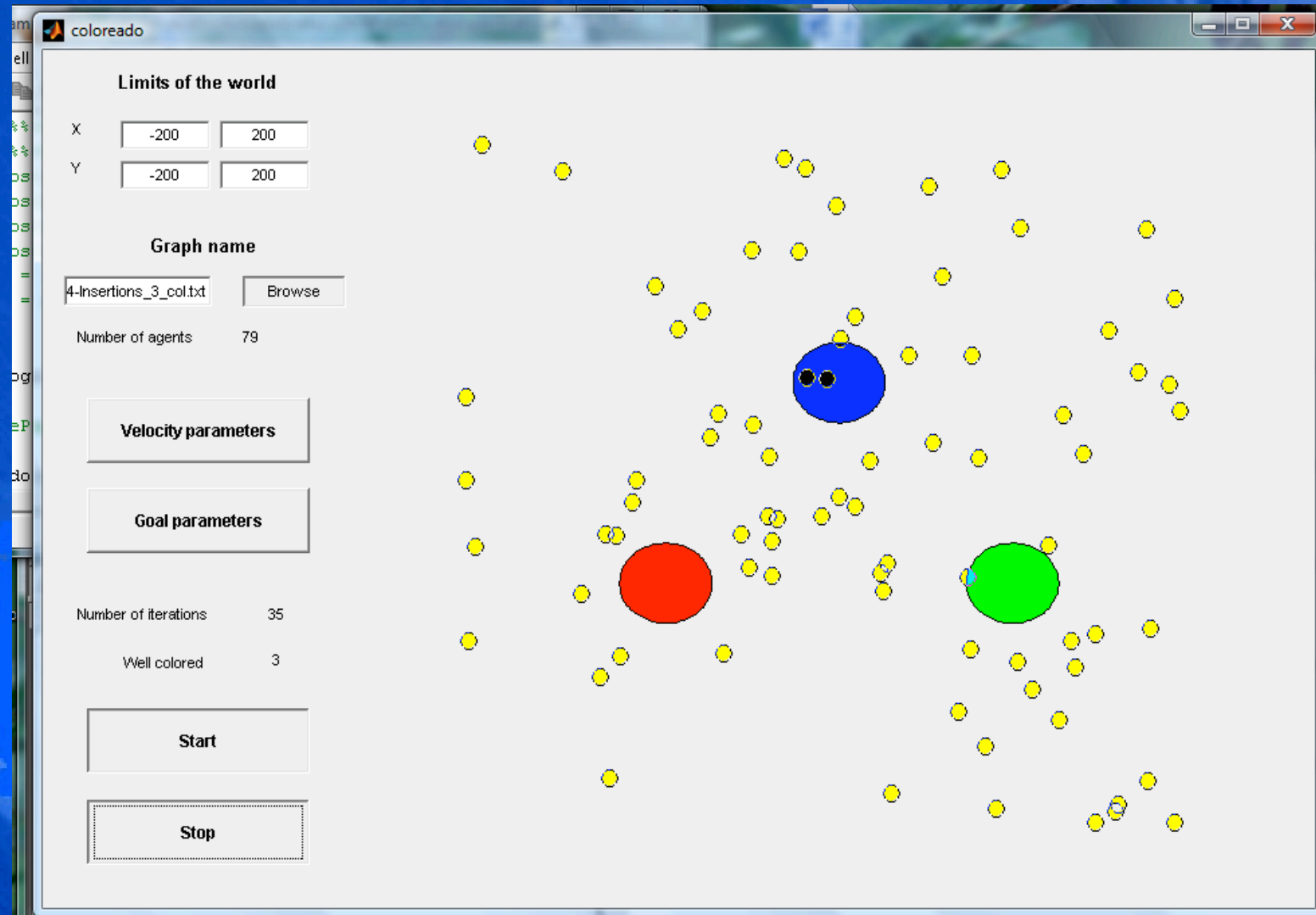
Looking for biological inspiration

- **General methods of optimization lose biological inspiration when representing some combinatorial problems.**
- **Steering behaviors are the simplest approach to swarm intelligence.**
 - Perception: position and velocity.
 - Action: change position.

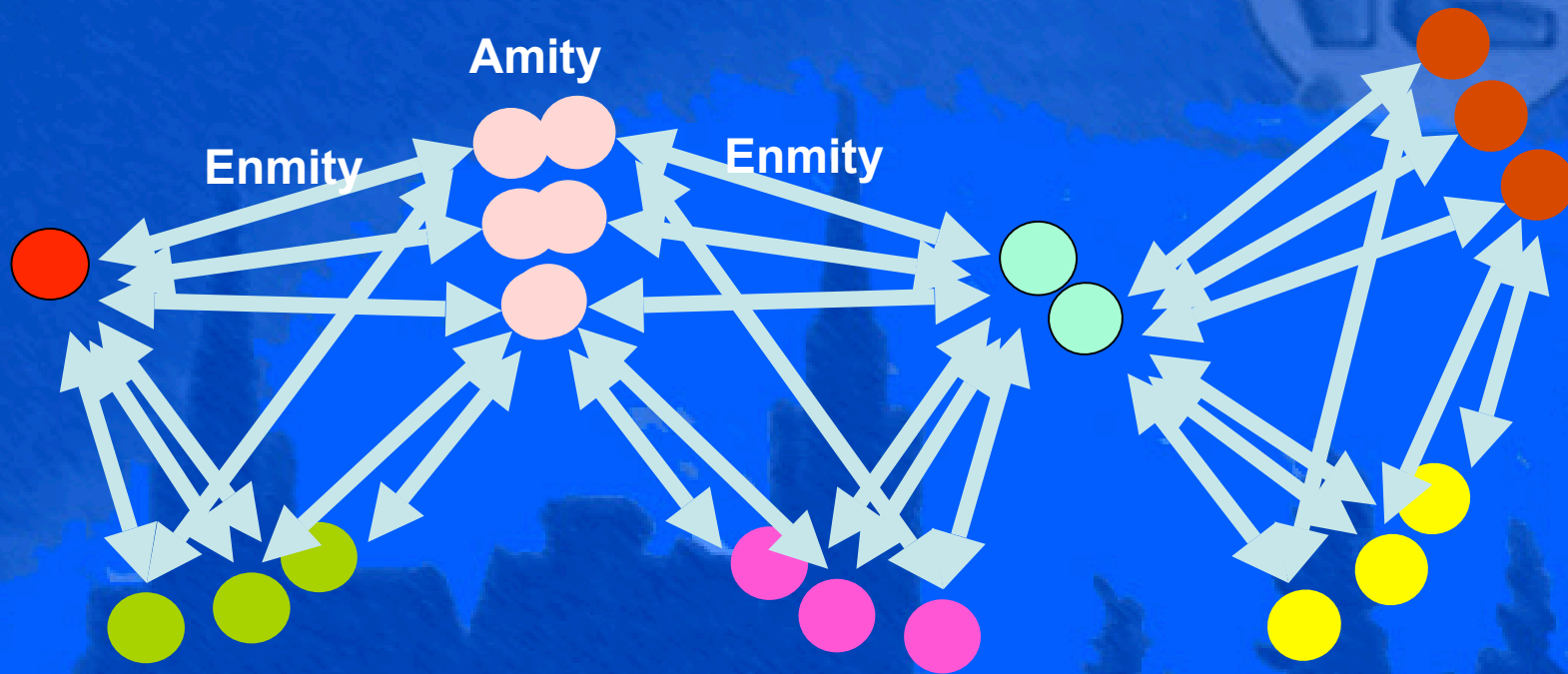
Hypothesis

- **Steering behaviors could be useful to find graph drawing algorithms that solve the problem of k-coloration:**
 - Toric 2D world.
 - Agents are navigating nodes.
 - Each color is a goal that attracts agents.
- **The steering behaviors of pursue-evasion are enough to represent coloration problem:**
 - Edges represent the relation of enmity.
 - The enemies of the enemies are friends.
 - An agent evades enemies and pursues friends.
- **Attack as a mechanism to avoid suboptimal solutions and ensure convergence if a solution does exist.**
- **For the sake of simplicity, center in $k=3$.**
- **The SOPS algorithm proposed will improve the correctness of the results obtained by Brélaz algorithm.**

Matlab application



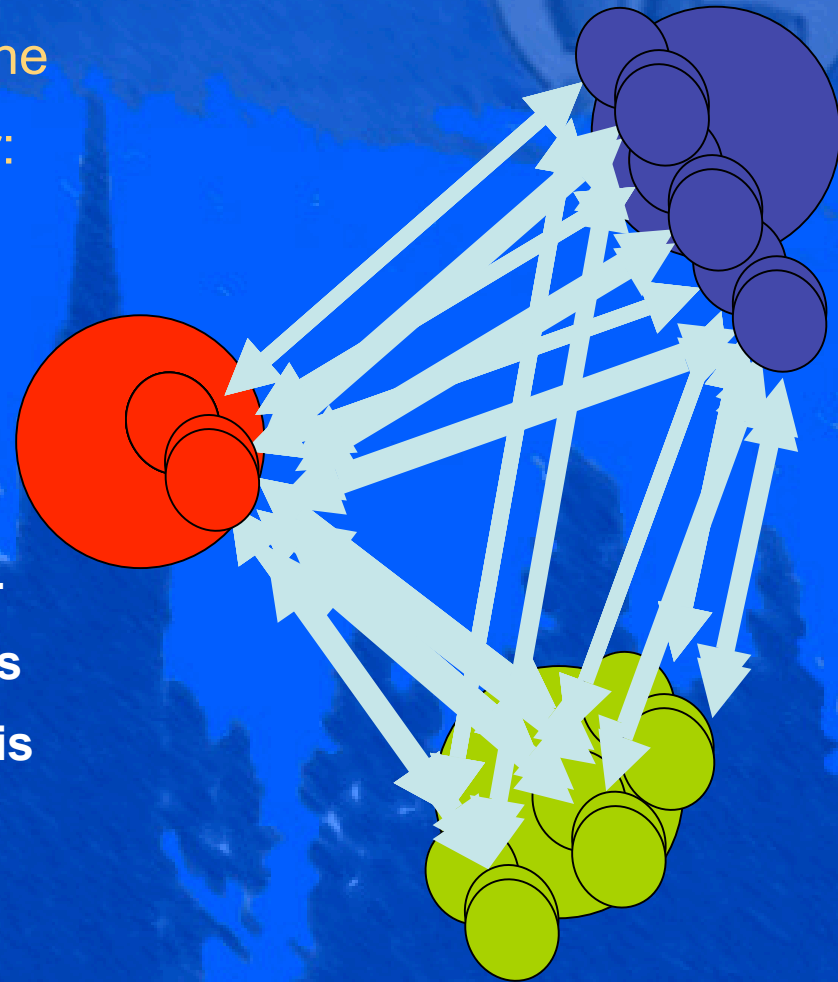
The dynamics of pursue-evasion:



- **Components: separation, cohesion and alignment**
- **If each cluster were a color obtains a consistent coloration**
- **More clusters than the chromatic number of the graph**

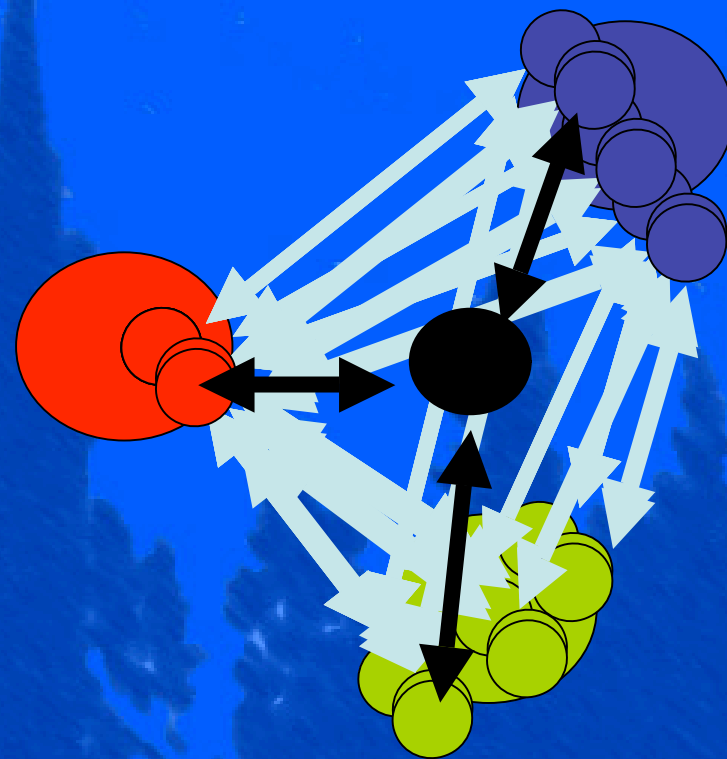
Seek toward goals

- Without the attack mechanism the system always converges either:
 - To an optimal configuration.
 - To a sub-optimal one:
 - with few nodes wandering around of the nearest goal.
 - with few nodes in the middle.
 - A sub-optimal configuration is reached whenever the graph is non 3-colorable



Attack: flee the goals

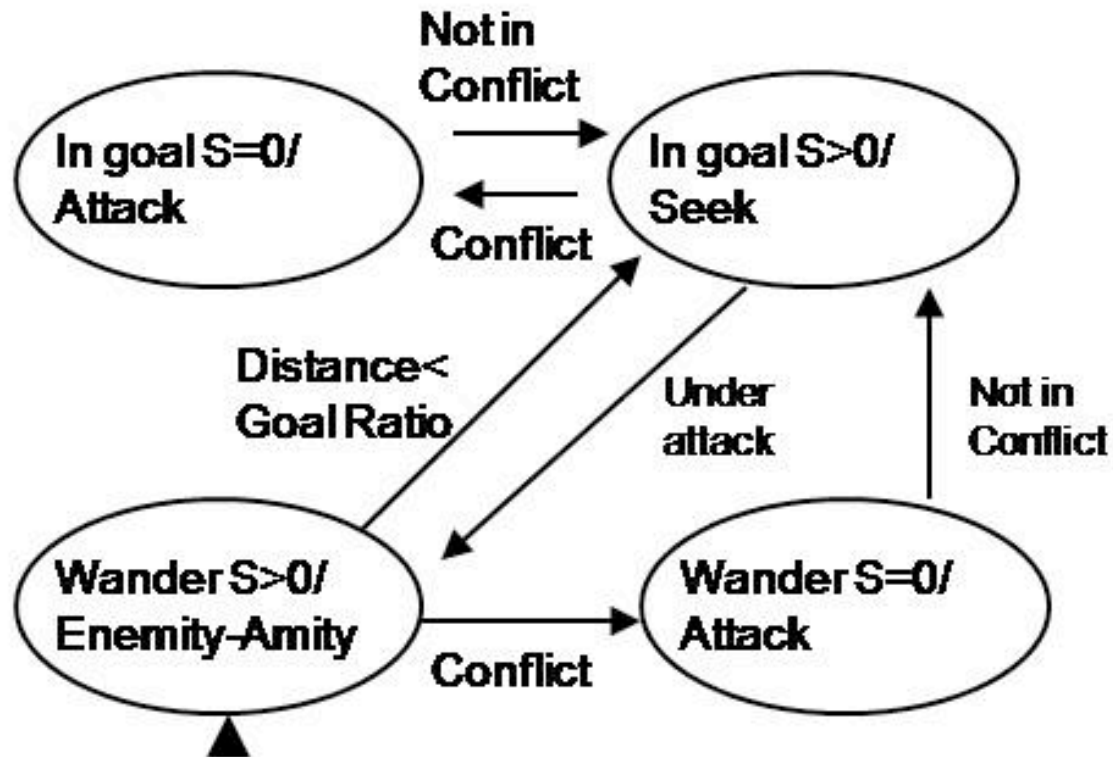
- **Internal conflict:**
 - Enemies in the same goal
- **External conflict:**
 - Enemies in all the goals



The rule of attack

- An internal counter of the degree of “desperation” or “dissatisfaction” of the agent in a conflictive situation.
- Agents in a goal have an increasing degree of satisfaction over time.
- In a conflict, the satisfaction level decreases until the counter reaches a value below a given threshold.
- In this case, the aggressive behavior is activated and the node attacks.
- The attack consists in selecting randomly an enemy in conflict which is less desperate than the aggressor.
- The node under attack is expelled from the goal and the aggressor takes its place.
- We have introduced a noise term in the velocity that helps to generate mildly erratic trajectories for wandering agents.

The FSM of individuals



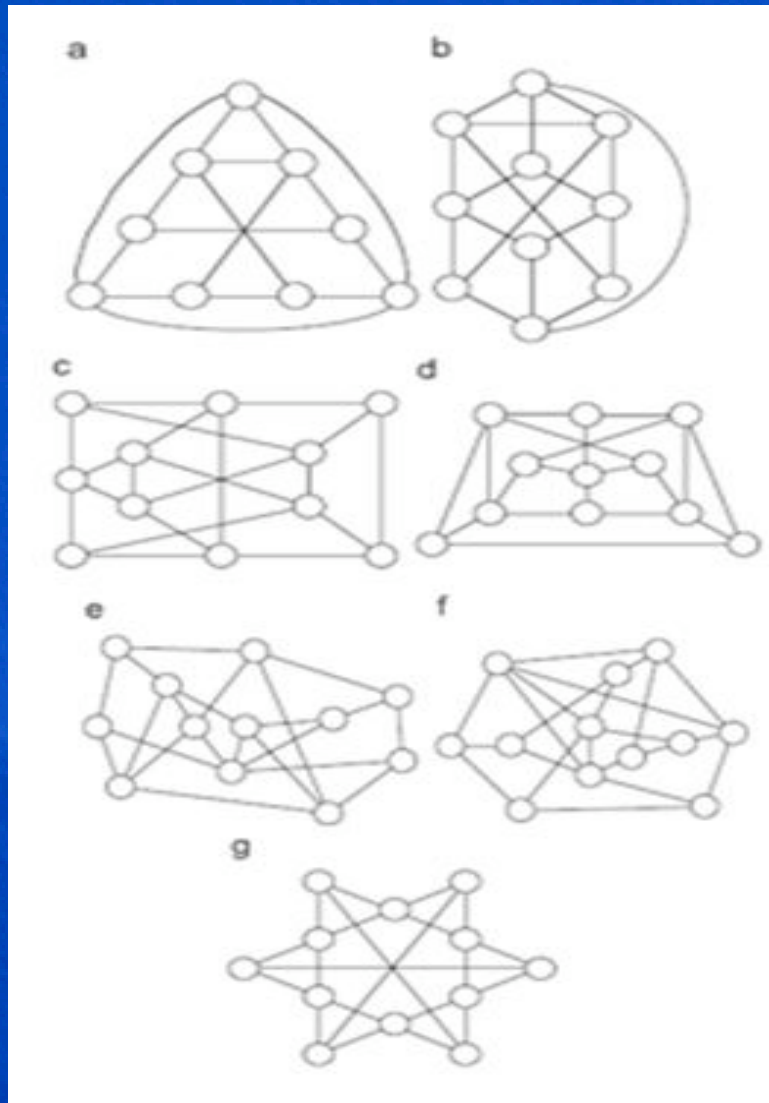
Benchmarking Experiments: a comparison to Bréaz heuristic

- The problem of 3-coloring of graphs has a very simple formulation but it is very difficult to solve.
- Conjecture of Steinberg (1979) open: every planar graph without 4 and 5-cycles is 3-colorable.
- We have made some experiments to verify that SOPS algorithm is at least as precise as Bréaz algorithm
- Meaning that the chromatic number given by SOPS is less or equal than Bréaz chromatic number.

Obtaining the sample

- **Good results for benchmark experiments.**
- **There is a class of graphs that are hard for 3-coloration by Brélaz algorithm**
Mizuno, K. and Nishihara, S. (2008).
- **Constructive generation of very hard 3-colorability instances.**
- **Algorithm implemented in Wolfram Mathematica.**

Hard graphs for 3-coloration



- **Embedding: combine graphs to obtain new ones.**
- **The grades of all nodes are 3 or 4.**
- **For all of them the Brélaz chromatic number is 4 while all of them are 3-colorable**
- **We generated a sample of 100 graphs**
- **Each graph a mean of 110 nodes by 10 random embeddings from basic graphs.**

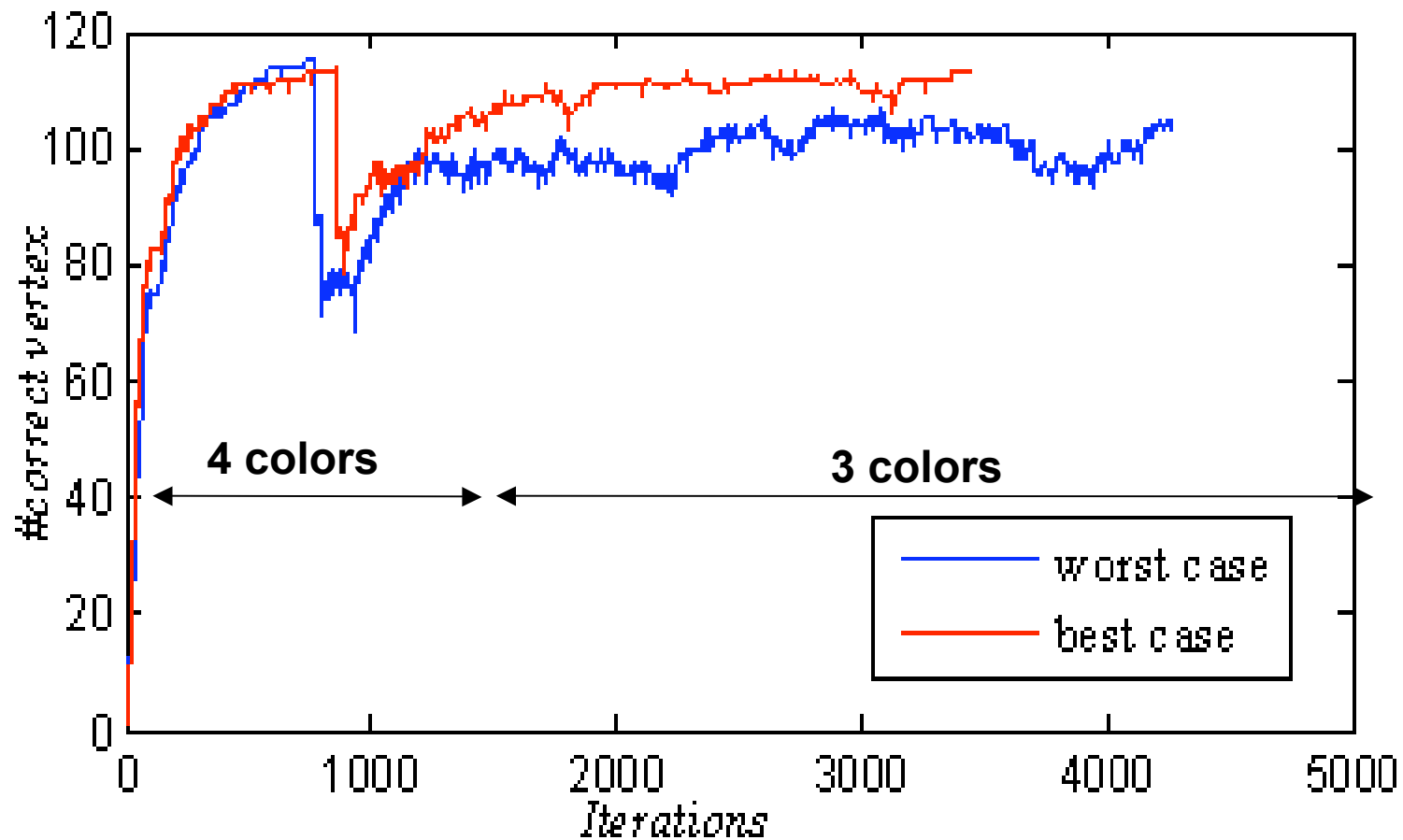
Experimental constraints

- **An experiment consists in:**
 - 25 runs of one graph.
 - **Registering the best configuration for each run:** the one that has minimum number of individuals in conflict.
 - Each run ends either when a 3-coloring solution is reached (success).
 - or when 5000 iterations are completed in cascade.
 - The number of iterations is registered.
- **Parameters are set in the default values.**

Cascade coloration strategy

- To obtain a faster convergence.
- The execution of the program has two stages :
 - First, the system attempts to find a 4-coloration (maximum 1500 iterations)
 - Second, eliminate the less populated goal. The individuals newly freed wander to seek a new goal
- until a 3-coloring is reached or the limit number of iterations (in this case 3500) are completed.
- This procedure of cascading coloration is based on known works in reaction-diffusion particle systems (Turk, 1991)
- To find the chromatic number of a graph, is sufficient to start the process of coloring successively the graph with colors $k, k-1, \dots$

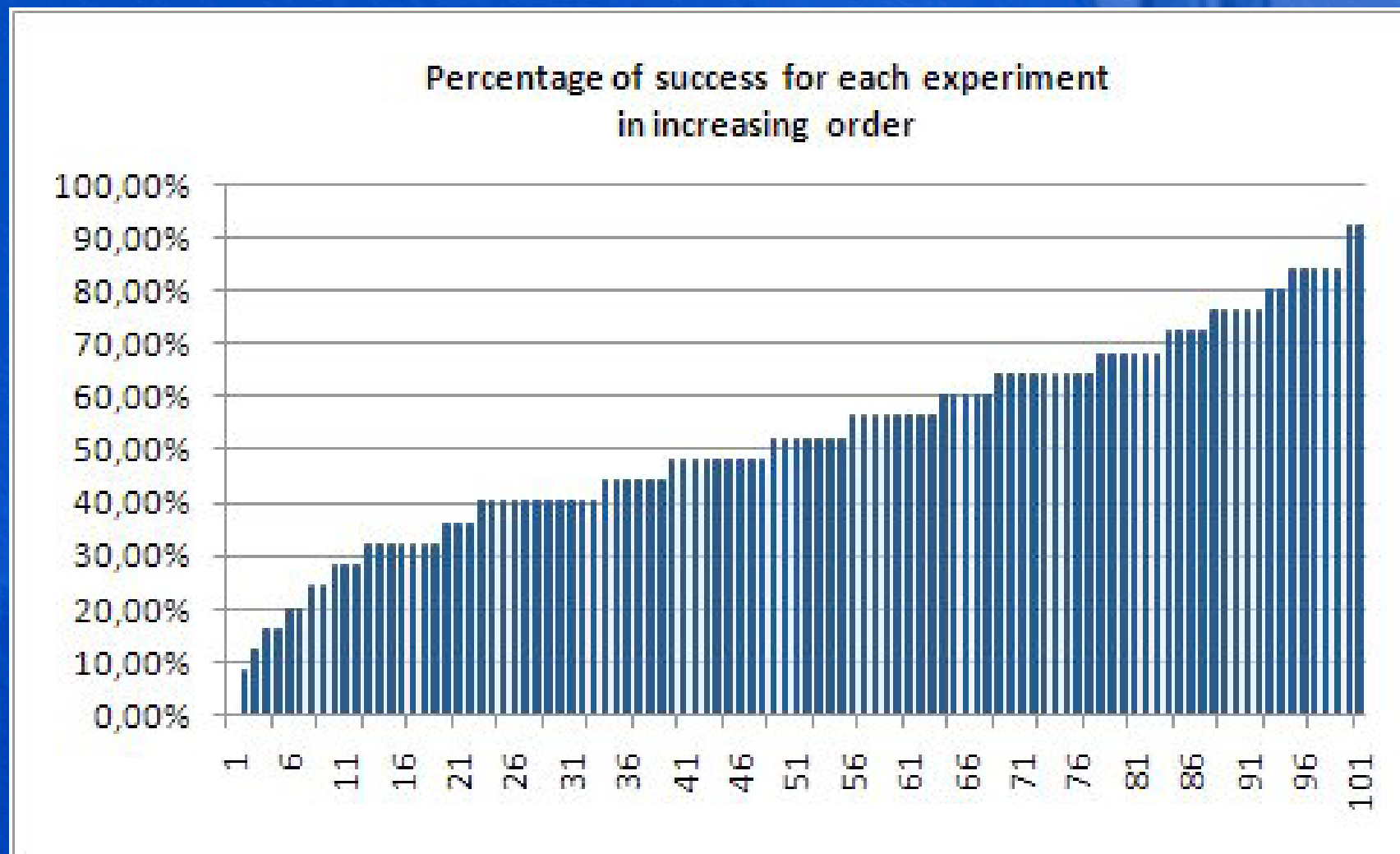
Number of correct colored vertices in cascade iteration of one experiment



Results over $100 \times 25 = 2500$ runs

- Mean number of boids: 110
- Mean number of iterations: 3761
- Average of succeeding runs: 51%

Success of SOPS 3-coloring

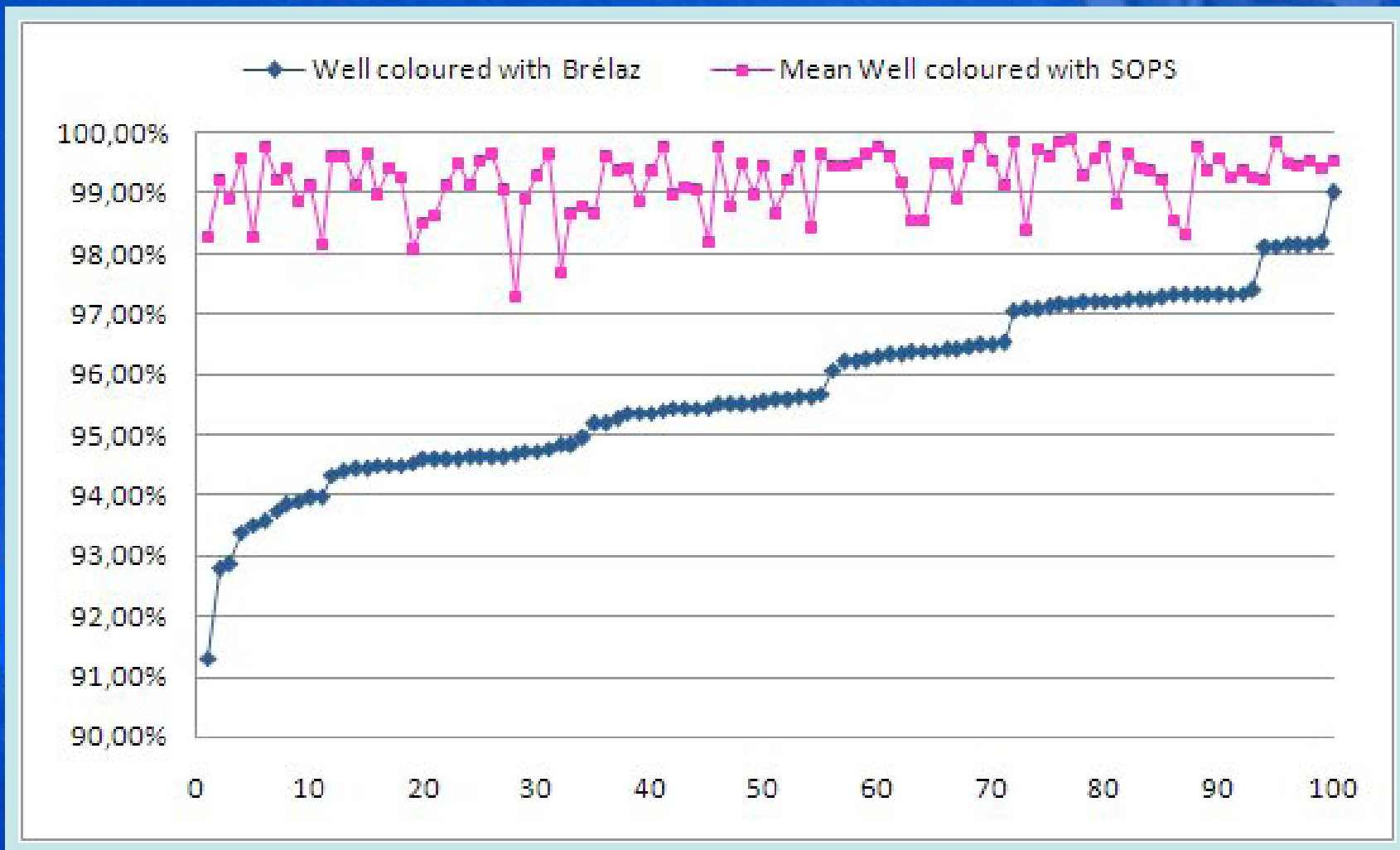


Counting errors



- Brélaz algorithm gives 4 colors for the 100 samples.
- Errors in Brélaz coloration: percentage of nodes of color 4.
- Errors in SOPS coloration: Take the best configuration through an iteration.
 - Minimum percentage of conflicting nodes.
- Average results over 100 experiments:
 - Brélaz algorithm colors well the 95,82% of the nodes with a standard deviation of 1.45%,
 - SOPS reaches a mean of the 99.17% and standard deviation 0.60%.
- Pearson coefficient of 0.30 has been found and in consequence, correlation does not exist between the results.

Brélaz versus SOPS for the benchmark hard graphs



Conclusions



- Self-Organizing Particle System solving the graph coloring problem.
- Cascading procedure of coloration makes the extension to k -colorations be an immediate consequence.
- We chose the problem of 3-coloring graphs because of the important open questions around the problem:
 - Is NP-complete
 - Steinberg's conjecture is giving an important research
- Other biologically inspired approaches to this problem:
 - (Dowsland and Thompson, 2008), ant colony optimization approach. identifies an individual in a population to a whole coloration of the graph, losing the biological inspiration in favor of cognitive abstraction.
- We do not know of any other attempt to solve the problem using flocking birds.

Conclusions

- Our approach to the coloration of graphs is mainly geometrical: graph drawing.
- The solution to the graph coloring emerges from the whole population configuration, which means a great economy of representation, and of computational power needed to implement the approach.
- The geometrical approach can be a source of experimentation and inspiration to improve sequential algorithms and heuristics for 3-coloration, which is important from the point of NP-completeness.

Questions?



Examples of the Matlab application

- fpsol2 i 1 colGRAPH.wmv: benchmark problem, 496 nodes and 11654 vertices, 0 fails.
- PetersenGraph.wmv. 10 nodes, 0 edges, 0 fails.

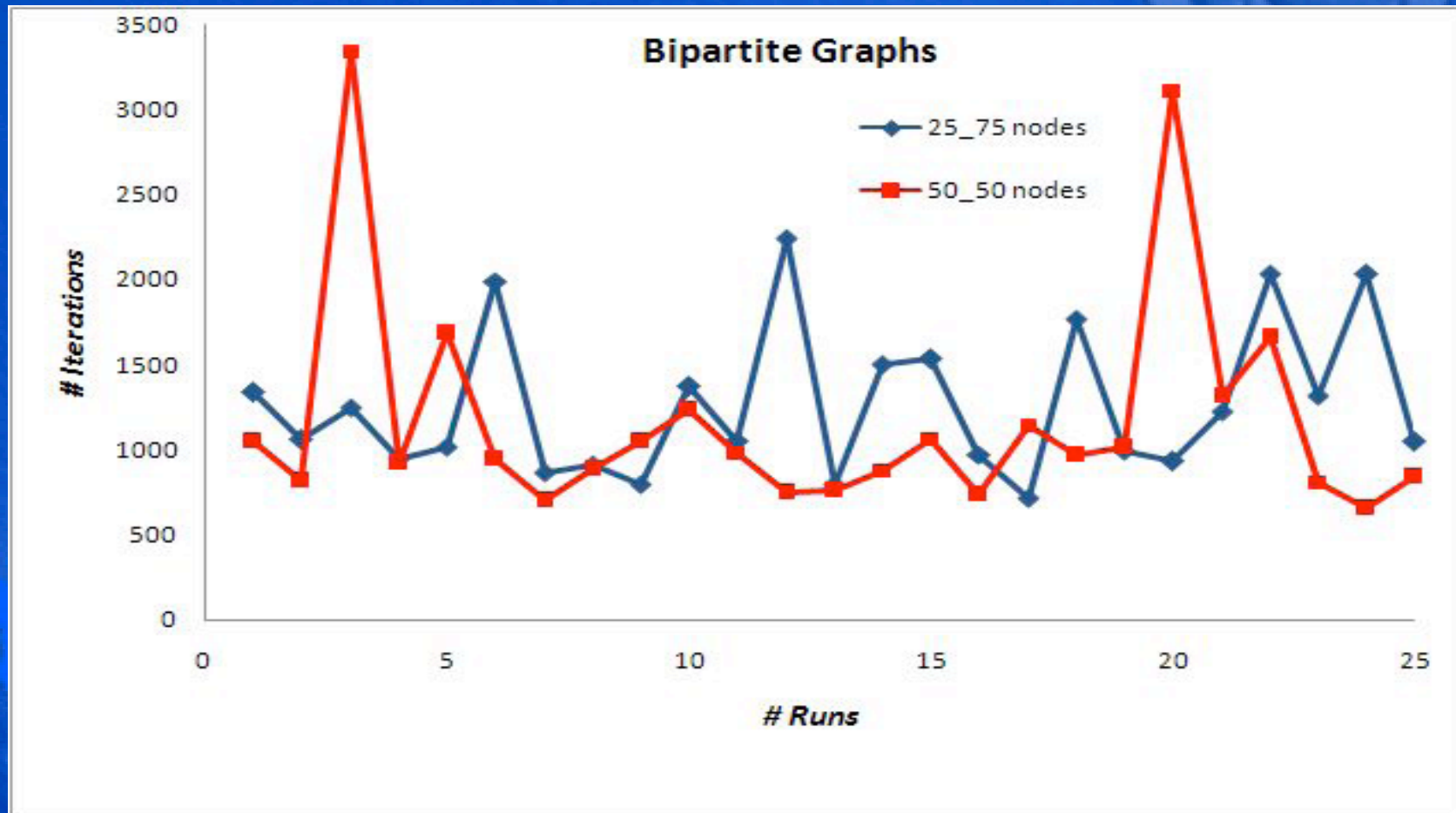
Results over $100 \times 25 = 2500$ runs

- Mean number of boids: $110 = n$
 $o(n) \leq \text{Complexity of one step} \leq o(n^2)$
- Mean number of iterations: 3761
 $o(n) \leq 3761 \approx 34 \times 110 \leq o(n^2)$
- Average of succeeding runs: 51%
One of each two runs is successful
 $n^3 = n^2 \times n < \text{Order of complexity} < 2 \times n^2 \times n^2 = n^4$

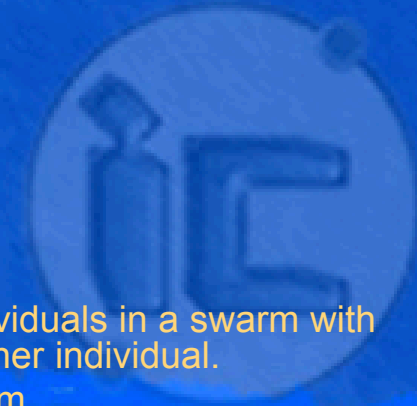
Percentage of well colored nodes

- It is well known that Brélaz algorithm needs two colors for a bipartite graph, being particularly efficient in this case.
- SOPS solves also correctly two complete bipartite graphs of 100 elements: 50-50 nodes and 25-75.
- In the 25 runs of each graph, the run was successful in both cases the 100% of the times.
- Regarding computing time measures, the mean number of iterations
- for graph 25-75 was 1266 and the minimum length of a successful run was 715.
- For graph 50-50 the average final step was 1172 being the minimum 654.

Iterations for 2 colorable graphs



Discussion



- Our aim was the research of the behavior arising from endowing the individuals in a swarm with another elementary cognitive ability: the perception of the affinity of another individual.
- The individual perceives another individual as belonging to We or to Them.
- We found that amity- enmity dynamics allows to model the solving process for coloring graphs.
- Complexity of swarms, understood as the complexity of the behavior of the emergent super-organism with respect to the computational capabilities of individuals. This work has been made in the last years in the field of theoretical computer science (Csuhaĵ-Varĵu et al., 1994; Kelemen and Kelemenov, 1992; Kelemenov´a and Csuhaĵ-Varĵu, 1994). We have attempted to discover the lowest computational capabilities of individuals that allows the swarm to perform a coloration of a graph. Revisiting the work of Rodriguez and Reggia (2004) may lead a strong theoretical basis for furthers developments in the convergence with grammar systems.
- The experimental results on hard coloring graphs with known chromatic number 3, show that the proposed approach can be very effective and competitive with state of the art algorithms. The Br´elaz algorithm algorithm is the common benchmark algorithm. Our approach improves on it over a sample of hard graphs.

Discussion



- We have designed and implemented a Self-Organizing Particle
- System that may interpreted as solving the graph colouring
- problem. We addressed the problem of 3-coloration of
- graphs, but the cascading procedure of coloration presented
- before makes the extension to k-colorations be an immediate
- consequence. We chose the problem of 3-coloring graphs
- because of the important open questions around the problem:
- it is NP-complete and Steinbergs conjecture is giving
- arise an important research nowadays (Borodin et al., 2005).
- A recent biologically inspired approaches to this problem
- has used the ant colony optimisation approach (Dowsland
- and Thompson, 2008), but we do not know of any other attempt
- to solve the problem using flocking birds. Their approach
- that identifies an individual in a population to a whole
- coloration of the graph, that is a tuple $(u_1; \dots; u_n)$ where u_i
- is the colour of node i , losing in this way the biological inspiration
- if favour of cognitive abstraction. On the other hand,
- our approach to the coloration of graphs is mainly geometrical,
- attending to the representation of the nodes of a graph
- as a flocking bird situated geographically. The solution to
- the graph coloring emerges from the whole population configuration,
- which means a great economy of representation,
- and of computational power needed to implement the approach.
- The geometrical approach can be a source of experimentation
- and inspiration to improve sequential algorithms
- and heuristics for 3-coloration, which is important from the
- point of NP-completeness.
- Second, we do not proceed in the direction of creating a
- model of colouring graphs from an existing model. Our aim
- was the research of the behaviour arising from endowing the
- individuals in a swarm with another elementary cognitive
- ability: the perception of the affinity of another individual.
- The individual perceives another individual as belonging to
- We or to Them. The first class is attractive while the second
- is repulsive. The first class is associated with amity, security
- and comfort while the second is interpreted as danger,
- enemies and things to avoid . We found that amity- enemy
- dynamics allows to model the solving process for coloring
- graphs, and not the other way around.
- The third contribution of this paper has to do with the
- complexity of swarms, understood as the complexity of the
- behaviour of the emergent super-organism with respect to
- the computational capabilities of individuals. This work has

Brélaz Algorithm



- Greedy algorithm for graph coloring.
- Colors={1, 2, ..., k}
- $D(v)$ =degree of node v = number of neighbors.
- $S(v)$ =saturation=number of different colors for neighbors.
 - Start in a vertex of maximum degree with color 1.
 - Color first the nodes v with maximum $S(v)$
 - Color node with maximum $D(v)$ if saturation is the same.
 - Use the minimum available color.
- Brélaz is a function provided by Wolfram Mathematica.

Basic rules of Reynold's model of boids

- **Separation:** steer to avoid crowding local flockmates.

$$v_s = - \sum_{b_j \in \partial_i} (p_j - p_i)$$

- **Cohesion:** steer to move toward the average position c_i of local flockmates

$$v_c = c_i - p_i \text{ where } c_i = \frac{1}{|\partial_i|} \sum_{b_j \in \partial_i} p_j$$

- **Alignment:** steer in the direction of the average heading of local flockmates.

$$v_a = \frac{1}{|\partial_i|} \sum_{b_j \in \partial_i} v_j - v_i$$

Seek and flee toward goals

- **Seek and flee:** seek attempts to steer a vehicle so that it moves toward a static goal. Here $\|p\|$ denotes the norm of a position or vector p and $f_{\text{maxvelocity}}$ is a non-negative parameter that limits the norm (the length in the Euclidean distance) of vector v_{seek} .

$$v_{\text{seek}} = v_{\text{goal}} - v_i \text{ where}$$

$$v_{\text{goal}} = \frac{p_i - p_0}{\|p_i - p_0\|} \times f_{\text{maxvelocity}}$$

Flee velocity is defined simply as the opposite of seek,

$$v_{\text{flee}} = -v_{\text{seek}}$$