



Ain Shams University
College of Women for
Arts, Science and Education
Physics Department

CONTRIBUTIONS TO DATA MINING USING ANT COLONY OPTIMIZATION ALGORITHMS

By

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Thesis Submitted
For
The Degree of Doctor of Science
(Physics – Computer Science and its Applications)

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University College of Women for Arts, Science and Education
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قسم الطبيعة

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ABSTRACT

In the past few decades, cluster analysis has played a central role in a variety of fields ranging from engineering (machine learning, artificial intelligence, pattern recognition, mechanical engineering, electrical engineering), computer sciences (web mining, spatial database analysis, textual document collection, image segmentation), to bioinformatics and medical sciences (genetics, biology, microbiology, paleontology, psychiatry, pathology). Clustering is often used as a tool for preliminary and descriptive data analysis and for unsupervised classification. Its main purpose is to identify homogeneous groups by finding similarities between objects regarding their characterizing attributes.

Swarm Intelligence (SI) is an innovative distributed intelligent paradigm for solving optimization problems that originally took its inspiration from the biological examples by swarming, flocking and herding phenomena in vertebrates. Particle Swarm Optimization (PSO) incorporates swarming behaviors observed in flocks of birds, schools of fish, or swarms of bees. Ant Colony Optimization (ACO) deals with artificial systems that are inspired from the

foraging behavior of real ants, which are used to solve discrete optimization problems.

Data mining and *Swarm intelligence* may seem that they do not have many properties in common. However, recent studies suggest that they can be used together for several real world data mining problems especially when other methods would be too expensive or difficult to implement.

This thesis deals with the application of swarm intelligence methodologies in data mining especially for data clustering. In our thesis PSO_Ant_Clustering algorithm represents the proposed clustering technique in which Ant Based Clustering is integrated by Particle Swarm Optimization for data clustering. The PSO_Ant_Clustering algorithm allows the parameters controlling the clustering sensitivity to be adapted as the PSO search algorithm proceeds. The proposed hybrid model consists of a set of particles, each of which has an Ant_clustering parameters set. The ants use these parameters to guide its search for favorable dropping/picking locations. The main goal of this is to create clusters of higher quality.

We have examined different types of problems. Using the traditional clustering techniques (such as K-Means and

Expectation Maximization), traditional Ant Based Clustering and the proposed technique PSO_Ant_Clustering algorithm are applied to the problems at hand. It has been shown that the PSO_Ant_Clustering algorithm has superior results than that of all such other techniques.

LIST OF PUBLICATIONS

1. El-Telbany M., Refat S., Abdelwhab A., Hefny H., and El Dakroury A., "*From Evolution to Swarm Intelligence: Data Clustering Survey*", International Journal of Intelligent Computing and Information Sciences, vol. 7, No. 2, pp. 219-236, July 2007.
2. El-Telbany M., Refat S., Sultan H. Aljahdali , "*Natural_Inspired Data Clustering: A Hybrization Between Ant Clustering and Particle Swarm Optimization*", in ISCA 18th International Conference on Software Engineering and Data Engineering, Las Vegas , USA, June 2009
3. Refat S., El-Telbany M., Abdelwhab A., El Dakroury A., and Hefny H., "Integrated Ant Based Clustering with Particle Swarm Optimization for Data Clustering", International Journal of Intelligent Computing and Information Sciences, vol 10., No 1. , pp. 291-302 , Jan 2010.

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