



**DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING
DELHI TECHNOLOGICAL UNIVERSITY**

DECLARATION

I hereby declare that the project entitled “**APPLICATION OF ARTIFICIAL BEE COLONY ALGORITHM USING HADOOP**” submitted by me in the partial fulfillment of the requirements for the award of the degree of Master of Technology (Software Engineering) of Delhi Technological University is record of my own work carried under the supervision and guidance of **Dr. Kapil Sharma**.

To the best of my knowledge this project has not been submitted to Delhi Technological University or any other University or Institute for the award of any degree.

SANJAY KUMAR
[2K13/SWE/20]

ACKNOWLEDGEMENT

In the sense of great pleasure and satisfaction I present this project entitled “**APPLICATION OF ARTIFICIAL BEE COLONY ALGORITHM USING HADOOP**”.

The completion of this project is no doubt a product of invaluable support and contribution of number of people.

I would like to express my sincere thanks to my guide **Dr. Kapil Sharma**(Associate Professor, Department of Computer Science and Engineering) for his continuous help and valuable suggestions and also providing encouraging environment, without which my project and its documentation would not have been possible.

The completion of any task is not only the reward to the person activity involved in accomplishing it, but also the person involved in inspiring and guiding. I am grateful to my friends and family for their constant motivation and comments that has helped me to complete this report.

SANJAY KUMAR

[2K13/SWE/20]

ABSTRACT

Optimization problem is a very common problem of finding the best solution from all feasible solutions. It is also a very common problem both in theoretical and practical scenes. In most cases optimization problem is also been classified as a NP-Hard problem. In today's era, time and computing resources are most common limitations. People often choose such metaheuristic methods that evaluate the optimal value in a limited time and with a limited computing resource. Artificial Bee Colony (ABC) is newly proposed metaheuristic algorithm which is suggested very compromising in recent work. As compared with the other metaheuristic algorithms, ABC is very effective for high dimensional problems. ABC can also solve various optimization problems with complex nonlinearity because it is a fast and robust algorithm. In this work, we have proposed a parallel implementation of Artificial Bee Colony algorithm for analysis of big data by using MapReduce paradigm. During the implementation of this work, Hadoop has been used as the backend MapReduce platform. As the result shows our ABC algorithm is very effective and scalable.

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