A COMPARISON OF BAT AND FIREFLY ALGORITHM IN NEIGHBOURHOOD BASED COLLABORATIVE FILTERING

THESIS



By: Bayu Permana Sejati 17.61.0105

BACHELOR OF INFORMATICS
FACULTY OF COMPUTER SCIENCE
UNIVERSITAS AMIKOM YOGYAKARTA
2020

A COMPARISON OF BAT AND FIREFLY ALGORITHM IN NEIGHBOURHOOD BASED COLLABORATIVE FILTERING

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to fulfil the requirements for a Bachelor's degree in the Informatics study program



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DECLARATION

I, the undersigned bellow, states that, this thesis is my own work (ORIGINAL) and the contents of this thesis have never been applied by any other person to receive an academic degree at a certain education institution, and as far as I know, there are no works or thoughts which have been written and/or published by anyone, except those in writing which are listed in this manuscript and which are mentioned in the reference list.

Anything that applies to the manuscripts and works that have been made is my own responsibility.



MOTTO

"Vivat academia, Vivant professores"
-Franz Liszt-

"Experience is the prime everyone gives to their mistakes."

- Oscar Wilde

"From zero to 0,1"

-Bayu Permana Sejati-

DEDICATION

Praise Allah SWT, for the grace of a great deal of mercy and His goodness, so that the author can complete the study entitled A Comparison of Bat and Firefly Algorithm in Negihbourhood Based Collaborative Filtering. Appreciation and gratitude give to:

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The authors concluded that this work has not been flawless, either in terms of content or presentation. Ideas and positive suggestions are expected at the conclusion of this thesis. Recently, the writers hope that this research will offer valuable things to the reader and, in particular, to the researcher.



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ABSTRACT

A Recommender System is a program for knowledge filtering that allows users to recommend items to users in a customizable manner and offer recommendations about the items that they might like. Recently, researchers have used the heuristic algorithm in the traditional technique of recommender system based on collaborative filtering or content-based filtering. In this research, the author proposes to prescribe the quality of recommender system by comparing swarm intelligence techniques, i.e. But and Friefly Algorithm. Swarm Intelligence uses to determine the user weight that provides by Pearson Correlation Coefficient. The result of this study shows Swarm Intelligence (But and Firefly Algorithm) able to improve the quality of Recommender System by considering the active users, and the Firefly Algorithm performs slightly better rather than But Algorithm by 0,0198 for Mean Absolute Error computation.

Keywords: But Algorithm, Firefly Algorithm, Recommender System, Pearson
Correlation Coefficient, Collaborative Filtering.