

Fast and Robust Hybrid Particle Swarm Optimization TABU Search Association Rule Mining (HPSO-TS-ARM) Algorithm for Web Data Association Rule Mining (WDARM)

Sukhjot Kaur¹, Monica Goyal²

Research Scholar¹, Assistant Professor²

Department of Computer Science and Engineering

Guru Kashi university, Talwandi Sabo

Punjab – India

ABSTRACT

Web search portals contains large amounts of web search data which includes keywords, links and other information. Web data association rules algorithm/s is the technique to deal with the web search data to produce the best results by analyzing the information in various combinations. In this paper, a novel web data association rule mining based hybrid algorithm called HPSO-TS-ARM has been proposed. This algorithms is based three well known high-level procedures: Particle Swarm Optimization, Tabu Search and Apriori Algorithm for Association Rule Mining. Where PSO will fetch the web search data in its optimized form, which is further computed by Tabu Search to prepare balance data arrangement followed by Association rule mining on processed web search data. The proposed algorithms have outperformed HBSO-TS and BSO-ARM on the basis of elapsed time and fitness function.

Keywords:- PSO, Bio-Inspired Algorithm, Association Rule Mining, Apriori, Hybrid Algorithm

I. INTRODUCTION

From recent couple of year explosive growth in amount of information or data come into notice. Data could be simple numerical figures, multimedia data, web data, text data and spatial data, is also being stored in files databases and data repositories. Finding ubiquitous model in a large amount of data is one of the key problems. For this particular reason data mining is attracted by information business and the world and is required to turn data into useful information and knowledge. Data mining is the process of fetching the desired information from large databases. Extracted information is used for different areas like business analysis, client maintenance, identifying the frauds, and scientific discoveries [1].

There exist diverse models of data mining such as classification, clustering, decision tree and neural networks from which association rule mining is also an important model. Association rules are used to extract the frequent patterns or casual structure among the set of items from given database. The pattern and rule discovered are based on the majority of commonly repeated items in dataset. Nowadays Association rule mining is broadly used in many different areas such as telecommunication networks,

market and risk management, inventory control mobile mining, graph mining, educational mining, etc. The traditional application of association rule mining is market basket analysis that considers the buying habits of customers. Market basket analysis also examines that how the items are purchased by the customer. Typical example of super market with large number of transactions for association rule mining is:

bread \rightarrow jam [sup=10%, conf=80%],

10% support states that of customer purchase bread and jam simultaneously, and 80% confidence means 80% customers purchased bread also buy the jam.

Formal statement of association rule mining is defined as:

Let $I = \{i_1, i_2, \dots, i_m\}$ be a set of items. Let T be a set of transactions (the database), where each transaction t (a data case) is a set of items such that $t \subseteq I$. An association rule is an implication of the form, $X \rightarrow Y$, where $X \subseteq I$, $Y \subseteq I$, and $X \cap Y = \emptyset$. The rule $X \rightarrow Y$ holds in the transaction set T with confidence c if $c\%$ of transactions in T that supports X also supports Y . The rule has support s in T if $s\%$ of the transactions in T contains $X \cup Y$.

Support and confidence are two basic parameters of association rules to generate the interesting association rules. To discover the interesting association rules domain experts specifies the minimum support (minsup) and minimum confidence (minconf) from given set of transactions item sets are called interesting if have greater support and confidence from minsup and minconf. To mine the association rules firstly find all the item sets having specified threshold support, secondly generate association rules from these item sets [2].

II. LITERATURE SURVEY

Rakesh Aggarwal et.al (1993) shows an algorithm that generates all significant association rules between items in database. The algorithm includes buffer management and new estimation and pruning techniques. Experimental result shows the effectiveness by applying to large retailing company [3].

Bing Liu et.al (1999) proposes a novel technique to solve the rare item problem to resolve the combinatorial explosion. Proposed model allows the user to specify the multiple minimum supports to show the nature of items and their varied frequencies in database and found rare item rules without producing meaningless rules with frequent items [2].

Habiba Drias et.al (2010) designed Bees Swarm Optimization algorithm named BSO-IR to explore the excessive number of documents to find the information desired by user. Experimental results shows better quality and runtime are compared between BSO and exact algorithms [4].

Y. Djenouri et.al (2012) proposes two new Association Rule Mining algorithms based on Genetic Meta-heuristic and Bees Swarm Optimization. Classical algorithms are not capable to cope with large data in lesser respond time. Proposed model achieves better while compared with IARMGA, AGA in both fitness criterion and CPU time [5].

Youcef Djenouri et.al (2013) proposes a novel hybrid algorithm HBSO-TS. It based on two meta-heuristics that are Bees Swarm optimization and Tabu Search. Experimental result shows better results as comparing to traditional approaches. They also planned the computation on GPU [6].

Mohammed J. Zaki et.al (1999) surveys the state of the art in parallel and distributed association-rule-mining algorithms and uncovers the field's challenges and open research problems also exposes that lot of exciting work remains to be done in system design, implementation, and deployment [7].

Takeshi Fukuda et.al (1996) discusses the data mining based on association rules for two numeric

attributes and one Boolean attribute. They consider two classes of regions, rectangles and admissible (i.e. connected and x-monotone) regions. They had implemented algorithms for admissible regions, and constructed a system for visualizing the rules [8].

David Martens et.al (2010) surveys two popular domains: swarm intelligence and data mining. Data mining has been a popular academic topic for decades; swarm intelligence is new subfield of artificial intelligence, based on social behaviour that can be observed in nature, such as ant colonies, flocks of birds, fish schools and bee hives. Framework that categorizes the swarm intelligence based data mining algorithms into two approaches: effective search and data organizing [9].

Dervis Karaboga et.al (2009) surveyed the algorithms based on intelligence in bee swarms and their applications. And surveyed algorithm VBA, ABC, BA developed for numerical problems can be expanded for combinatorial types problems by suitable modifications [10].

III. PROPOSED WORK

The existing web data association rule mining algorithms HBSO-TS and BSO-ARM yield good results and are adequate performers than the other comparative algorithms. But there is always a possibility of improvement in the existing algorithms. When we studied the case of the HBSO-TS and BSO-ARM, it was found that there is a lot of possibility to create an improved algorithm than the existing ones using the combination of PSO and ARM algorithms which are known as Particle Swarm Optimization and Association Rule Mining respectively. In this research, we are proposing the new possible solution which will be better performer than the existing ones in terms of elapsed time and fitness function. We have proposed a new web data association rule mining algorithm with improved results than the existing ones. The new algorithm is called HPSO-ARM which is based on a hybridization of Particle Swarm Optimization and Association Rule Mining. The three major components of the proposed algorithm are the particle swarm optimization and association rule mining. Particle Swarm is a bio-inspired algorithm used to find the solution in a given space or area by computing the possible formation of the particles in that give space. The computation run until the best solution is found. To perform web data association rule mining, we have used apriori algorithm. Apriori algorithm is primarily used with transactional databases. This algorithm analyzes the individual entities in the database and extends them to find the item/entity sets. These item sets are called frequent item sets, which are further used to

determine to association rules to find the general trends in the database. Hence, this algorithm is easily adaptable to web data association rule mining.

IV. METHODOLOGY

We have proposed a new web data association rule mining algorithm with improved results than the existing ones. The new algorithm is called HPSO-TS-ARM which is based on a hybridization of Particle Swarm Optimization, Tabu Search and Association Rule Mining. The three major components of the proposed algorithm are the particle swarm optimization, tabu search and association rule mining.

Particle Swarm Optimization: As mentioned earlier, Particle Swarm Optimization simulates the activities of bird’s flocking. Assume that a faction of birds are erratically searching foodstuff in a particular area. Only single piece of foodstuff is existed in the area, which is being searched by the birds. All of the birds in the group are not aware about the actual position of the food. Hence, the question is “what is the best strategy to find the food?” The most useful technique is to find the bird closest to the foodstuff. PSO cultured from the latter scenario is used to solve such optimization problems. Every solution is the bird in the area of interest or area of search. It is called a particle. Each particle carries a fitness value which is evaluated by the fitness function. The birds have different velocities which direct the flight of the particles. The particles take flight through the area of interest by succeeding the existing optimal particles. PSO is initialized with a group of random particles or solutions, followed by the search for optimal value or optima by updating generations. Every solution is rationalized by finding two best values. First is optimal solution or fitness. It is also called pbest. Second values is followed by the particle swarm optimizer is the second best value, attained by any particle in the bird population. It is called global best or gbest. When computations are performed on a particle taking part in the population as per its neighbors in the topology, best values becomes local best of lbest.

After obtain the optimal values, the solution updates velocity and positions of itself with following equation(i)and(ii).

$$(i) \quad v [] = v [] + CL1 * rand() * (pbest[] - present[]) + CL2 * rand() * (gbest[] - present[])$$

$$(ii) \quad present[] = present[] + v []$$

Where v [] denotes the particle velocity, present[] denotes current solution. Pbest [] and gbest [] are fitness and global best respectively. Rand () is a

random number between (0,1). CL1, CL2 are learning factors.

Pseudo code of the particle swarm optimization is explained as following:

Algorithm 1: The general Algorithm PSO

1. For each particle
2. Initialize particle
3. END
4. Do
5. For each particle
6. Calculate fitness
7. If fitness value is better than pBest in history
8. set pBest equal to current value
9. End
10. Choose the pbest of all the particles as the gBest for all particle
11. Calculate particle velocity according equation (a)
12. Update particle position according equation (b)
13. End

Tabu Search: The main idea behind tabu search, which is a local search metaheuristic to rearrange the data in a particular manner to find the most effective solution called tabu list. Tabu search finds the best neighbors to maintain the tabu list L. The process runs the maximum number of iterations for a given condition to find the best solutions for the input data.

Algorithm 2 The General Algorithm TS

1. S ← Some Initial candidate solution.
 - a. Best ← S.
2. L ← { } a tabu list of maximum length 1.
3. I ← 1.
4. **while** i < Max-Iter and not stop **do**
5. Enqueue S into L.
6. S ← Best neighbors(S).
7. Alter Best if Quality (Best) < Quality(S).
8. **end while**

Apriori- Association Rule Mining: To perform web data association rule mining, we have used apriori algorithm. Apriori algorithm is primarily used with transactional databases. This algorithm analyzes the individual entities in the database and extends them to find the item/entity sets. These item sets are called frequent item sets, which are further used to determine to association rules to find the general trends in the database. Hence, this algorithm is easily adaptable to web data association rule mining.

Algorithm 2 The General Algorithm TS

Data set	HBSO--TS	BSO-ARM	HPSO-TS-ARM
Bolts	1.0	1.0	1.0
Sleep	1.0	1.0	1.0
Pollution	1.0	1.0	1.0
Basket Ball	0.97	0.97	1.0

C_k : Candidate item set of size k
 L_k : frequent item set of size k
 $L_1 = \{\text{frequent items}\};$
for ($k = 1; L_k \neq \emptyset; k++$) **do**
 - C_{k+1} = candidates generated from L_k ;
 - **for each** transaction t in database **do** increment the count of all candidates in C_{k+1} that are contained in t ;
End for;
 - L_{k+1} = candidates in C_{k+1} with min_support
End for;
return $\cup_k L_k$;

V. RESULT ANALYSIS

In our proposed algorithm HPSO-TS-ARM algorithm is developed using Particle Swarm Optimization, Tabu Search and ARM are combined together. It was a difficult task to choose the datasets and performance parameters. Various databases are tested under this research and the results are obtained in the form of elapsed time, elapsed time graph and fitness value. The result analysis confirms that HPSO-TS-ARM algorithm outperforms the HBSO-TS and BSO-ARM. We have performed all of the results on the the MATLAB v2011a installed on Windows PC with i3 processor and 2GBs of RAM. The PC was also equipped with 1GB GPU, but it was not utilized for the experimental computations. The proposed algorithm is taking almost half or less than half of the time than the existing HBSO-TS and BSO-ARM algorithm for web data association.

The datasets tested under this research project are Bolts, Sleep, Pollution and Basketball. These datasets

Data set Name	Transaction Size	Item Size
Bolts	40	8
Sleep	56	8
Pollution	60	16
Basket Ball	96	5

are used because they were compact in size and the algorithm we developed is most applicable on the compact sized datasets. According to our experimental design analysis, we found that this

algorithm will also perform better on the large sized datasets.

Table 1: DATA SETS DESCRIPTION

Table 2: Our Approach to Other Approaches W.R.T. To Fitness

Dataset Name	HPSO-TS-ARM (time in seconds)
Pollution	27.7265
Bolt	12.3406
Basketball	15.0894
Sleep	19.0366

Table 3: ELAPSED TIME BY HPSO-TS-ARM

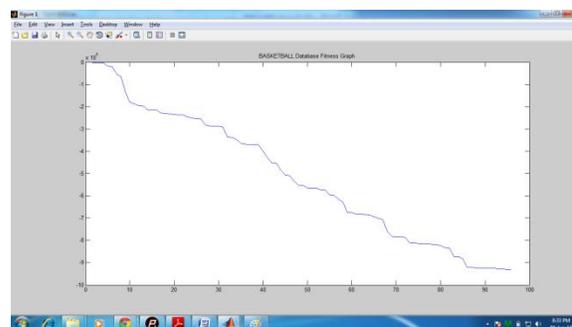


Fig. 1: Basketball DB Fitness Function

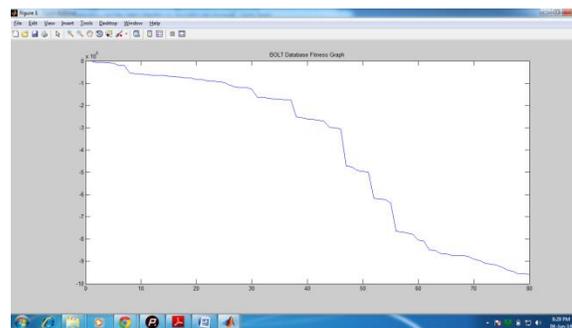


Fig. 2: Bolt DB Fitness Function

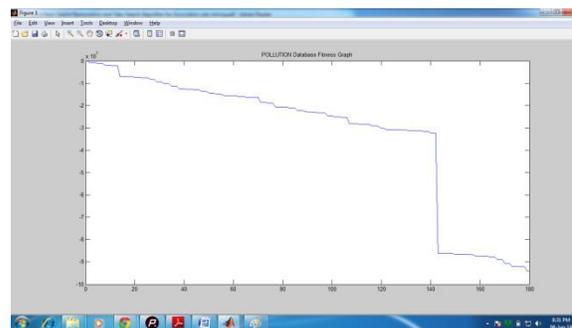


Fig. 3: Pollution DB Fitness Function

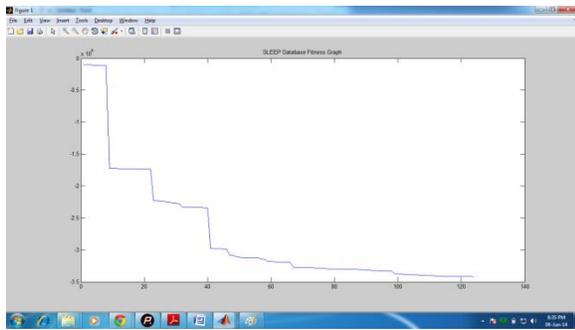


Fig. 4: Sleep DB Fitness Function

VI. CONCLUSION

The proposed algorithm named as HPSO-TS-ARM algorithm. HPSO-TS-ARM has equally performed or outperformed the existing HBSO-TS and BSO-ARM algorithms in terms of fitness value. The performance parameter of elapsed time has been tested on the latter mentioned four databases which yields good and acceptable results. An equal or better fitness function values with less elapsed time is a sign that proposed algorithm will also perform better results on large datasets. In future, this algorithm will be tested with more datasets and will be compared with the HBSO-TS and BSO-ARM in the terms of other performance parameters also. Its performance will be also tested and compared with other similar algorithms on the basis of various datasets and more performance parameters. Because the proposed algorithm is proved to be useful for the web data association rule mining, it will be enhanced to perform better than the proposed algorithm by combining it with different algorithms to develop new algorithms using new algorithmic combinations or newly developed algorithms.

REFERENCES

- [1] Han, J., Kamber, M., & Pei, J., Data mining: concepts and techniques. *Morgan kaufmann*. 2006.
- [2] Liu, B., Hsu, W. and Ma, Y., "Mining association rules with multiple minimum supports", *KDD '99 Proceedings of the fifth ACM SIGKDD international conference on Knowledge discovery and data mining*, ISBN: 1-58113-143-7, pp 337-341, August 1999.
- [3] R. Agrawal, T. Imielinski and A. Swami, "Mining association rules between sets of items in large databases", *SIGMOD '93 Proceedings of the 1993 ACM SIGMOD international conference on Management of data*, ISBN: 0-89791-592-5, Volume 22 Issue 2, pp 207-216, May 1993.
- [4]. Habiba Drias, Hadia Mosteghanemi, "Bees Swarm Optimization based Approach for Web Information Retrieval", *International Conference on Web Intelligence and Intelligent Agent Technology IEEE/WIC/ACM*, ISBN: 978-0-7695-4191-4, Volume 1, pp 6-13, September 2010.
- [5]. Y. Djenouri, H. Drias, Z. Habbas, H. Mosteghanemi, "Bees Swarm Optimization for Web Association Rule Mining", *International Conferences on Web Intelligence and Intelligent Agent Technology, IEEE/WIC/ACM*, ISBN: 978-1-4673-6057-9, Volume 3, pp 142-146, December 2012.
- [6] Youcef Djenouri, Habiba Drias, Amine Chemchem, "A Hybrid Bees Swarm Optimization and Tabu Search Algorithm for Association Rule Mining", *World Congress on Nature and Biologically Inspired Computing (NaBIC) IEEE*, ISBN: 978-1-4799-1414-2, pp 120-125, August 2013.
- [7] Mohammed J. Zaki, "Parallel and Distributed Association Mining: A Survey", *Concurrency IEEE*, ISSN: 1092-3063, Volume 7, Issue 4, pp 14-25, December 1999.
- [8] Takeshi Fukuda, Yasuhiko Morimoto, Shinichi Morishita, Takeshi Tokuyama, "Data Mining Using Two-Dimensional Optimized Association Rules: Scheme, Algorithms, and Visualization", *ACM SIGMOD international conference on Management of data*, Volume 25 Issue 2, pp 13-23, June 1996.
- [9] David Martens, Bart Baesens, Tom Fawcett, "Editorial survey: swarm intelligence for data mining", *Machine Learning Springer*, ISSN: 0885-6125, Volume 82, Issue 1, pp 1-42, January 2011.
- [10] Dervis Karaboga, Bahriye Akay, "A survey: algorithms simulating bee swarm intelligence", *Artificial Intelligence Review Springer*, Volume 31, Issue 1-4, pp 61 – 85, June 2009.