Improved Memetic Algorithm Enabled Intelligent Multi Agent (IMAEIMA) System for Web Mining

D. Weslin, T. Joshva Devadas

Abstract: Wider web space, the searching of a relevant data is the most curious problem for the common people accessing the web. For retrieving the relevant information the user request is given to search engine. The relevant pages combined with irrelevant pages are returned to the user. The proposed work emphasizes an Improved Memetic Algorithm Enabled Intelligent Multi Agent (IMAEIMA) for searching the most appropriate pages when submitting complex queries. Improved Memetic algorithm is the traditional genetic algorithm combined with local search and random selection. In this proposed system Improved Memetic algorithm additionally enhanced with logarithmic weight function for more accuracy. Intelligent Agents are introduced in this IMAEIMA to improve its performance and accuracy by reacting intelligently based on feedback and previous experience. This system helps to retrieve relevant pages from web databases with high precision and recall. The derived architecture reveals greater precision and recall overriding the conventional search algorithms.

Index Terms: Genetic algorithm, Intelligent Agent, Memetic Algorithm (MA), Web Mining.

I. INTRODUCTION

Recent years World Wide Web (WWW) has agressive growth and thereby experiences several obstructions in retrieving the relevant information from the available data set. Recent trends like Semantic Web (SW), multi-agent systems, data mining, offers knowledge beyond information retrieval from the the Web. For that, various techniques have been proposed for retrieving the most accurate pages for the submitted complex query. Improper phrase or sentence with more number of keywords is termed as complex query. The search engine searches the relevant pages by choping the query into keywords. Most of the pages are retrieved by web crawler. These pages are may be relevant or irrelevant for results. In the proposed system, user query can be processed by the Improved Memetic Algorithm Enabled Intelligent Multi Agent (IMAEIMA) system and finds the relevant pages from the web. This system receives query from the user and checks the page existence in the knowledge data base (KDB), if exists then the relavant pages are immediately sent to the user. If the page doesnot exist in the knowledge data base (KDB) the Process Agent processes the query by using search engine for relevant snippets and thus be processed by Improved Memetic Algorithm for the most relevant links.

II. RELATED WORKS

Gatial et al. (2013) proposed a system to collect the information and also integrates by semi-automated agent technique. Agent binds the methods for design and generation of form to end user [1]. Aarti Singh, (2012) focused on clustering techniques for mining information from the web based on agent [2]. The web information is grouped into clusters relevant to the query; satisfying user needs and yields better usage of web surfing time [2]. Shakti Kundu et al, (2017) introduced a new system called WEBMINTEL, is a multi agent system with three offers: reject, accept or add to cart, and these offers are intelligently executed based on prior knowledge [3]. Melita et al. (2012) reviewed the characteristics of Genetic Algorithms, GA based search techniques, problems in implementation and available methods suitable for web search optimization. Further, they suggested success rate improvements using enhancements in further research [4]. Kolli et al. (2013) proposed a new method to find the optimal patterns from the web. This technique combines Apriori algorithm and GA operators with association rule mining approach [9]. The unrequired patterns are removed and the target patterns are extracted from the frequent patterns [9]. Raval Pratiksha et al. (2014) constructed dual clusters with greater consistence between the users and pages [5]. At last the most visited pages by all its users are retrieved. The users who react as similarly on subset of pages are listed [5]. Chuang et al. (2014) proposed a web crawling approach for the extraction of automatic
address and associated information for positioning the location entity on maps [6]. Das et al. (2015) presented a new technique for identifying the query forms from webpages [7]. Their technique exhibits high accuracy in the extraction of the desired data [7]. Pooja Solanki et al. (2015) recommended a system to predict the user’s navigational behaviour using the biclustering approach with greedy search and genetic algorithm to overcome the problems of grouping available in traditional clustering approach [8]. Jayaprabha et al. (2016) exploded the possibility of genetic algorithm for extracting relevant information from the web, using fitness function and ranking function with genetic algorithm [10]. The parameters for this method are extracted from the web usage and web content mining with better results [10]. Panjwani Heena et al. (2017) system used coherent dual clustering framework to trace consistent dual clusters using the click-stream information [12]. This framework can be applied to enhance web page design, information access and feature of service provided [12]. The objective of this work found high volume dual clusters with great consistency of the user and visited pages. Wang et al. (2017) suggested a crawling method based on the document frequency of the queries to mitigate the ranking bias problem [11]. Ferrante Neri et al. (2012) focused on characteristic of memetic structure particularly the coordination of memes [13]. Surekha More et al. (2014) proposed genetic algorithm based intelligent approach for web mining. Genetic Algorithm combined with local searching technique enhances the web page search [14]. The similarity between several snippets is computed and the most similar snippets are returned. Khushali D et al. (2015) proposed improved memetic algorithm using hybrid selection strategy with local search for enhancing the results [15].

III. IMAEIMA ARCHITECTURE

Nowadays, several search engines provide the necessary information from World Wide Web data bases, but the outcome need to be refined to reach the accuracy. While using complex queries the issue of getting extraneous results is observed. Query with many key words is termed as Complex query. The proposed Improved Memetic Algorithm Enabled Intelligent Multi Agent (IMAEIMA) system for web mining is based on intelligent agents coupled with Improved Memetic Algorithm (IMA). The proposed system has four intelligent agents such are:

1. User Agent (UA)
2. Process Agent (PA)
3. KDB Agent (KDBA)
4. Good Link Agent (GLA)

The user query enters into the Intelligent Multi Agent (IMAs) system, the agent immediately triggers the searching in the knowledge data base (KDB) and the relevant page exists then responds it to the user. The relevant page is not hit by agent then the agent sends the user request to the conventional search process. The conventional search engine provides the snippets for the submitted query. Successive searches are done after preprocessing the snippets and query. These preprocessed snippets constitute the first generation for Improved Memetic Algorithm (IMA). Heuristic-approach will be taken over for local search process. Apply Term frequency matrix, Weight matrix, cosine similarity and fitness function for generating the next generations. The GA processes of selection, crossover and mutation will be carried out. End of the process the system extracts the positive and passive samples automatically through Good Link agent, and updates the KDB with new snippets. The user receives the new relevant results. A greater enhancement is obtained in the search process using an Intelligent Multi Agent with Improved Memetic Algorithm (IMAEIMA). Fig.1. shows the architecture of IMAEIMA system

Procedure of proposed system:

1. User enters the Query Q.
2. Q received by Intelligent Multi Agent system (IMAs).
3. Intelligent Multi Agent system (IMAs) checks the Q with the KDB.
4. If Q exists in KDB then send it to user;
5. Else
6. Begin:
7. Send Q to Search engine
8. Apply Improved Memetic Algorithm (IMA) for further generations, using conventional search engine results.
9. After completion of Improved Memetic Algorithm (IMA) process, the results are checked by the Good Link agent.
10. The KDB updated by KDB agent and sends the results to user.
11. Stop the process.

A. Intelligent Multi Agents

An automated software component interacting in the system to replace human intervention is termed as agent. Based on prior experience the intelligence of the agent is trained to enhance the performance of the system. This intelligent agent finds vital role in innovative real world applications. The agents are characterized by self learning, based on knowledge and their previous experience added with knowledge updation. The proposed IMAEIMA system introduces four intelligent agents to improve the performance. These are User agent, Process agent, KDB update agent and Good Link agent. Fig.2 shows the communication in between agents.

User Agent (UA): User agent establishes the connection in between user and the system. It receives the Query from an user and sends the results the user. And also it communicates with KDB agent.

Process Agent (PA): Process agent communicates with search engine, KDB agent, IMA and Good Link agent. Process agent receives the Query from KDB agent and sends this Query to IMA for further generations. After this IMA process it receives the results from IMA and sends these results to Good Link agent.
**KDB Agent (KDBA):** KDB agent communicates with KDB, User agent, Process agent and Good Link agent. KDB agent receives the query from user agent and checks its availability in KDB, if exists reply to the user agent with results. Otherwise send the query to process agent for further process. And also collects the results from Good Link agent for KDB updation.

**Good Link Agent (GLA):** Good Link agent communicates with KDB agent and Process agent. Good Link agent receives the results from PA and checks the positive and passive results. These results send to KDB agent.

**B. Improved Memetic algorithm approach**

Memetic Algorithms are a population-based approach. Memetic algorithm is local search process embedded in evolutionary algorithm. GA taken for evolutionary process and heuristic function is used for local search process. Memetic algorithm is characterized by the same parameters involved in GA:

- Population count
- Generation count,
- Crossover probability
- Mutation probability and along with a local search process. In addition to enhance the results hybrid-selection process is incorporated with MA is called Improved Memetic Algorithm (IMA). In IMA the population is initialized using a heuristic function or random method. A local search technique is employed to compute fitness value of each individual. The individual possessing higher quality of...
fitness forms the next generation population. New individuals are generated by crossover and mutation operation performed between two selected parents. The aspect of local search in memetic algorithm is to arrive at the local optimum accurately with minimum number of iterations.

**Improved Memetic Algorithm Procedure:**

**Step 1:** Initial population from problem space.

**Step 2:** Preprocessing of snippets and query
- Eliminate the stop words and spaces.
- Query transformed into keywords.

**Step 3:** local search process based on heuristic function
\[ H(i) = \sum F_{ik} \]
where \( n \) is number of snippets in the population.

**Step 4:** Calculate Term Frequency matrix using \( K \) and Snippets, row \( i \) represents snippet \( i \) is a term frequency vector
\[ TF_i = \{TF_{i1}, TF_{i2}, TF_{i3}, \ldots, TF_{in}\} \]

The quality of snippet \( i \) is calculated as
\[ F(i) = \sum_{k \in K} TF_{ik} \]
The cosine similarity between snippet \( i \) and snippet \((j)\) is tabulated from
\[ s(i, j) = \frac{\sum_{k \in K} TF_{ik} \cdot TF_{jk}}{\sum_{k \in K} TF_{ik} \cdot \sum_{k \in K} TF_{jk}} \]

Where \( F_{ik} \) is the frequency if keyword \( k \) in snippet \( i \). The cosine similarity produces \( n \ x \ n \) matrix output for next generation of snippets.

From Term Frequency Matrix \( n \ x \ m \) (\( n \)-snippets, \( m \)-keywords in the query) generate weight matrix using, the keyword \( k \) in snippet \( i \).
\[ W_{k,si} = \begin{cases} \frac{1 + log_{10}(TF_{ik})}{max_{1 \leq i \leq n}(TF_{ik})} & \text{if} \; TF_{ik} > 0 \\ 0 & \text{otherwise} \end{cases} \]

From \( S(i,j) \) and \( W_{k,si} \) can be used for finding fitness function to measure the performance of snippet \( i \) and snippet \( j \), it rewards the next generation chromosomes.
\[ F(i, j) = \sum W_{k,si} \cdot s(i, j) \]

Where \( W_i \) is the weight of the \( Q \) in node \( i \) and \( W_q \) is the total number of query terms in \( Q \).

**Step 5:** After fitness computation the snippets are ranked to generate chromosomes passed on to next generation.

**Step 6:** After selection process the Crossover and Mutations are performed.

**Step 7:** Iterate step 3 to 7.

**Step 8:** Send the final results.

**B.1. Hybrid-Selection strategy**

In IMA the selection process follows hybrid strategy. Initially the random population is generated. The chromosomes are sorted in order in one pool and in random in the other pool. Fig.3, explains the hybrid selection strategy. The hybrid-selection order is indexed from 1 to the size of population, denoted by \( i \). From each set the chromosome at first position is selected. The hybrid strategy is retrieving best chromosome by tournament selection between the first chromosomes of each set \( P \) and \( RP \). Each chromosome participates in the selection process, so that the chromosome with least fitness value also survives. Crossover is done several times between the two parents so that the best characters of the parents are reflected in next generation. Normally in crossover operation the child is carried out to next generation avoiding the features of the parents. The least offspring fails to reach the local minima. In the proposed

**B.2. Cross over**

GA is an evolutionary algorithm used to build an optimum result by combining the chromosomes together based on its good characters. This system uses a single point cross over technique for offspring generation. In single point cross over technique one cross over point is selected, using this cross over point swapping segments of right side of two parents produce the two off springs. In the following example 5 th position is the cross over point. So, right sides of the two parents are swapped to get offsprings.

| Parent 1 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
|----------|---|---|---|---|---|---|---|---|
| Parent 2 | 5 | 6 | 3 | 1 | 7 | 8 | 2 | 4 |
| Child 1  | 1 | 2 | 3 | 4 | 7 | 8 | 2 | 4 |
| Child 2  | 5 | 6 | 3 | 1 | 5 | 6 | 7 | 8 |

**B.3. Mutation**

During mutation step, the selected individuals are used as parents to produce the children of the succeeding generation. The objective of mutation operator is maintaining diversity. The chromosomes in the next generation may be typical or atypical to the parent chromosomes. The chromosomes are discarded in the next generation, if they are atypical. The population progresses toward an optimal solution over the successive generations.

In this system order of changing technique is incorporated, in which two positions are selected and interchanged. In the below example 2nd and 7th elements are interchanged to generate offspring.

[5 6 3 1 5 6 7 8] → [5 7 3 1 5 6 6 8]
The proposed IMAEIMA architecture is an improvement over MA and IMA, which does not use an intelligent agent technology experience to retrieve the relevant information. In the proposed IMAEIMA architecture, the query is received by intelligent multi agent and checked with KDB for existence. After existence checking the user request is executed by conventional search process and first generation is computed. From this the term frequency is calculated using function (2). The quality of population is calculated using function (3). For local search process the heuristic function (1) is used, and select the population for further process. After, selection the fitness values of relevant documents to be calculated using fitness function (6) with help of weight function (5) and cosine similarity function (4). Based on fitness threshold value the best populations are processed using GA operations such as crossover and mutation. The new offsprings are added with existed population. This process continues till reach the threshold condition based on number of iterations. These outputs are received by the intelligent multi agent and are checked by the Good Link agent (GLA) for the positive and passive results. These results will be stored in KDB by KDB agent and simultaneously send it to user. Thus, use of intelligent multi agent reduces the over head of the system. In the IMA process the weight matrix for each keyword is calculated using logarithmic function. The logarithmic weight function helps to reduce the larger values into small values for easy and fast calculations. This provides more accuracy for finding fitness values. The local search with hybrid selection of the population and quality of population reduces the iterations.

The intelligent agent checks the final results by GLA once again reduce the error rate of non relevant documents. In information retrieval system the performance can be measured by Precision and Recall. The measure of documents relevant to search is termed as Precision. Recall is the measure of the documents that are relevant to the retrieved user request.

\[
\text{precision} = \frac{|\text{Relevant documents} \cap \text{Retrieved documents}|}{|\text{Retrieved documents}|}
\]

\[
\text{Recall} = \frac{|\text{Relevant documents} \cap \text{Retrieved documents}|}{|\text{Relevant documents}|}
\]

This system uses 10 complex queries and the cross rate is 0.5 and mutate rate is 0.01 with 17 iterations. The result values are plotted in F-score graph by the experience based on the significance of the retrieved pages. These results are compared with MA and IMA results.

Fig.5. Precision Vs. Recall Metric for IMAEIMA

Fig.6. Precision Vs. Recall Metric for IMAEIMA

The relevancy of each snippet is the measure of its precision and recall values. The plotted results of precision versus recall are depicted in the following Fig.4, Fig.5 and Fig.6 respectively. From these results the proposed IMAEIMA system is produced significantly better results.

V. CONCLUSION

Thus the implementation of a search engine using intelligent multi agents with improved memetic algorithm additionally enhanced with logarithmic weight function produces more accurate results. The multi agents are intelligently reacts in the environment based on its prior experience and its frequent feedback system. And also it reduces the error rate using Good Link Agent. In future the improvement of the system may be based on search time and also implementation of multi search engine instead of only one search engine and cross over combined from different search engine results.

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