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Integration of Cloud Computing and Big Data Technology in Computer Informatization Construction



Abstract: - The combination of big data with cloud computing technologies has surfaced as a paradigm-shifting approach in computer information building, fundamentally altering the ways in which enterprises handle, evaluate, and employ copious volumes of data. In this manuscript, Integration of Cloud Computing and Big Data Technology in Computer Informatization Construction (ICCBTD-CIC-RBAGCN-FOA) proposed. Initially input data are gathered from Internal Revenue Service Dataset. To execute this, input data is pre-processed, Adaptive-Noise Augmented Kalman Filter (ANAKF) and data cleaning; then pre-processed image is fed to computer information construction utilizing Relational Bilevel Aggregation Graph Convolutional Network to evaluate and analysis the process of computer information construction. In generally, RBAGCN doesn't express adapting optimization strategies to determine optimal parameters to ensure Cloud Computing and Big Data Technology based computer information construction. Therefore, Fox-inspired Optimization Algorithm (FOA) is to optimize Relational Bilevel Aggregation Graph Convolutional Network which accurately constructs computer information. Then the ICCBTD-CIC-RBAGCN-FOA is implemented in Python and performance metrics such as Accuracy, Precision, Recall, Acceleration rate, Task execution rate and Response Time are analysed. Performance of ICCBTD-CIC-RBAGCN-FOA approach attains 18.57%, 22.15% and 31.10% higher accuracy, 19.59%, 23.12% and 32.60% higher Precision and 19.57%, 25.15% and 31.60% higher Recall when analysed through existing techniques like Contactless technologies for smart cities: big data, IoT, and cloud infrastructures(CTSC-BDCI-SVML) , Designing an Accounting Information Management System Using Big Data and Cloud Technology(DAIMS-BDCT-RBM) , Construction of Digital Platform of Religious and Cultural Resources Using Deep Learning and Its Big Data Analysis(CDP-RCR-BDA-CNN) methods respectively.

Keywords: Adaptive-Noise Augmented Kalman Filter, Big Data Technology Cloud Computing, Fox--inspired Optimization Algorithm, Relational Bilevel Aggregation Graph Convolutional Network.

I. INTRODUCTION

As machine learning and artificial intelligence continue to merge, computers have become an indispensable tool for human productivity and daily life [1]. These development have been very important, especially in the field of data processing, where they permit for the storage of enormous volumes of data as well as the execution of statistics and analytics on the data until the usefulness of the data sources is further explored [2, 3]. Application databases have made information handling considerably easier and more accurate for people [4]. Data sources have therefore become more and more important in the fields of business, administration, and research. Since the year 2000, as computer technology has advanced and become more widely used, many facets of life have produced more data than they could have imagined while making significant advancements [5-7]. Because of the aforementioned, the term "info explosion" is frequently used to describe the exponential growth in data [8]. People's daily activities and lives have become more difficult as a result of the amount of information, especially when it comes to finding pertinent information quickly [9]. Modern firms can only thrive and expand by comprehending consumer expectations and market dynamics and providing goods and services that meet those prospects [10, 11]. Seeking relevant information is necessary in a highly competitive market to effectively employ data sources for market research, user monitoring, and analytical decision-making. [12]. Irrelevant information has become a hassle for organizations in the process of obtaining useful information, impacting not only the effectiveness of information processing but also functioning as a deceiver [13]. Data collection technology is therefore very valuable and contributes to significant future growth. Given that financial management is closely related to a company's ability to survive and grow, it is regarded as being crucial to enterprise management [14]. It is also an important measure for evaluating how well businesses are doing [15]. As network and information technology advance quickly, an increasing number of companies are concentrating on the informationization of financial management [16]. The combination of big data and cloud computing has become a key factor in the advancement of computer informatization construction in the age of digital transformation [17]. These technologies present previously unheard-of possibilities to advance enable data-driven decision-making, operational efficiency, and stimulate innovation across a range of industries as businesses work to handle and use massive volumes of data [18].

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Through the internet, on-demand access to a shared pool of computational resources, cloud computing offers scalable. It frees businesses from having to spend money on and maintain expensive infrastructure, enabling them to use computer power, storage, and services as needed [19]. Additionally, cloud systems are affordable, flexible, and agile, allowing businesses to quickly expand their operations and adjust to changing business conditions [20].

The design of an Accounting Information Management System utilizing Big Data, RBM and Cloud Technology may provide obstacles related to data security and privacy, as well as the intricacy of effectively connecting several systems. Building a Digital Platform for Religious and Cultural Resources with CNN and Big Data Analyses might have trouble correctly deciphering complex cultural settings. It might also have trouble guaranteeing inclusivity and respect for different religious perspectives, and it might have biases in its algorithmic interpretations. Implementing contactless technologies for smart cities with SVMML may face hurdles in ensuring real-time data accuracy and reliability, while also grappling with the scalability of IoT devices and potential security vulnerabilities in cloud infrastructures.

In this research, a novel method for improving groundwater level prediction ICCBDT-CIC-RBAGCN-FOA is presented. The system seeks to achieve high Accuracy and precision by combining FOA with optimized RBAGCN. Pre-processing methods is ANAKF. The suggested approach is put into practice using Python and contracted to the state-of-the-art techniques. It shows notable improvements in a number of measures, such as Accuracy, precision, Recall, Acceleration rate, Task execution rate and Response Time.

Below is a summary of this research work's primary contribution:

- The ICCBDT-CIC-RBAGCN-FOA model's resilience enables precise Computer Informatization Construction in Cloud Computing and Big Data Technology.
- Optimized Relational Bilevel Aggregation Graph Convolutional Network with Fox-inspired Optimization Algorithm use to enhancing parameter of RBAGCN.
- The ICCBDT-CIC-RBAGCN-FOA model's resilience enables precise computer information construction.
- The proposed method is put into practice on the Python platform, and its effectiveness is compared to that of current approaches.
- The proposed method yields superior outcomes when compared with other current techniques like DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML.

Remaining manuscripts arranged as below: segment 2 describes literature review; segment 3 depicts proposed method, segment 4 clarifies outcomes with discussions, segment 5 conclusion.

II. LITERATURE SURVEY

Many research works presented in the literatures were based on computer information construction based deep learning; few of them were reviewed here,

Dai, [21] has presented DAIMS-BDCT-RBM. This portion investigates the notion of cloud computing and looks at its logical structure in order to employ the model and technology in the design of an accounting information management system. Next, it builds the RBM model to construct the accounting information management systems (AIMS) cloud platform architecture. Once the system design and construction are complete, the system's application effect is assessed by evaluating the cluster performance and figuring out the cloud platform's distributed storage mode. Lastly, experimental tests were conducted for load balancing, local rows of data, and system operating time. It attains higher Accuracy and it provides higher Acceleration rate.

Sun, [22] has presented CDP-RCR-BDA-CNN. Here, a CNN based on clustering is developed using the PCA vector as the convolution kernel. In order to extract additional information, the CNN creates numerous sets of convolution kernels, clusters tiny images, and computes principal component vectors based on categories. This enables adaptive feature extractor selection for the input image. Additionally, it offers a DL-based technique for extracting picture features. A comprehensive investigation was conducted on the creation of a digital platform for religious and cultural material, along with its big data analysis. Additionally, an algorithm for DL-based image feature extraction was suggested. Here uses simulation and comparison analyses to confirm the effectiveness of the method. It attains higher Task execution rate and it provides higher Response Time.

Manimuthu et al. [23] have presented CTSC-BDCI-SVML. This article concentrates on technology-driven applications that might affect the municipal infrastructures now in place as they develop into contactless smart cities. Applications, design guidelines, technological standards, and affordable methods that make use of BIC for contactless applications were presented. User interfaces used in smart city settings were also covered. Here, also used over cutting-edge sensing techniques and intelligent software that enable smart contactless features in cities. Lastly, a case study detailing how BIC may help effectively handle and manage emergency situations was presented. It attains higher Recall and it provides lower Accuracy.

Narayanan et al. [24] have presented “novel system architecture for secure authentication and data sharing in cloud enabled Big Data Environment”. In this instance, data owners were registered with a trust centre through the use of SHA-3 hashing in big data outsourcing. Using the Map Reduce architecture, the input file was split up into fixed-size data blocks. The SALSA20 technique was then used to encrypt each block. Big data sharing involves people retrieving files in a safe manner. The user's credentials were hashed and matched with those kept in a database for this reason. Three crucial procedures were used in big data management to arrange data. They were as follows: Fractal Index Tree for indexing, DBSCAN for clustering, and Lemperl Ziv Markow Algorithm (LZMA) for compression. It achieves greater accuracy and quicker response times.

Zhang [25] has presented “Digital Transformation of Enterprise Finance under Big Data and Cloud Computing”. In this case, enterprise finance deals with additional needs and problems pertaining to firm financial management and control. In the real world, it's important to investigate how to assess enterprise financial management issues in light of the company's unique circumstances and how to successfully integrate big data and cloud computing technologies into enterprise financial control operations. It attains higher precision and it provides higher Acceleration rate.

Stergiou, et al [26] has presented “InFeMo: flexible big data management through a federated cloud system”. Here, suggested approach intends to expand the breadth of data management by providing users with a more energy-efficient system architecture and environment. Additionally, the user would spend less time waiting in each procedure queue by implementing the InFeMo. In order to better and more quickly manage user requests, the PaaS paradigm was used to build the proposed system on the resources made accessible by CSPs. The goal of presented method was to close a knowledge gap in the area of federated cloud systems. It attains higher precision and it provides lower Task execution rate.

Chen, et al [27] has presented “Enterprise financial management information system based on cloud computing in big data environment”. In order to find a solution for the cloud-based enterprise financial management information system in a big data environment, cloud computing technology can, in this case, significantly lower the application threshold of enterprise financial informatization construction, improve the informationization's return on investment, and flexibly adapt to the needs of different business stages. This is a result of its excellent dependability, flexible growth options, inexpensive investment costs, and on-demand services. The management idea of "business-driven value" was put forth in model as a method of controlling expenses. It attains higher precision and it provides lower recall.

III. PROPOSED METHODOLOGY

In this section, Optimized RBAGCN depend Informatization Construction Method for Computer Informatization Construction (ICCBDT-CIC-RBAGCN-FOA) is proposed. This process comprises of four steps like data collection, pre-processing, Construction, and optimization. In the proposed Computer Informatization Construction system, Internal Revenue Services Dataset undergoes pre-processing to prepare them for further analysis. Following pre-processing; the final step involves employing a RBAGCN for Computer Informatization Construction. The Fox--inspired Optimization Algorithm method is introduced for training the RBAGCN. Block diagram of ICCBDT-CIC-RBAGCN-FOA approach is represented in Fig 1. As a result, a thorough explanation of each step is provided below.

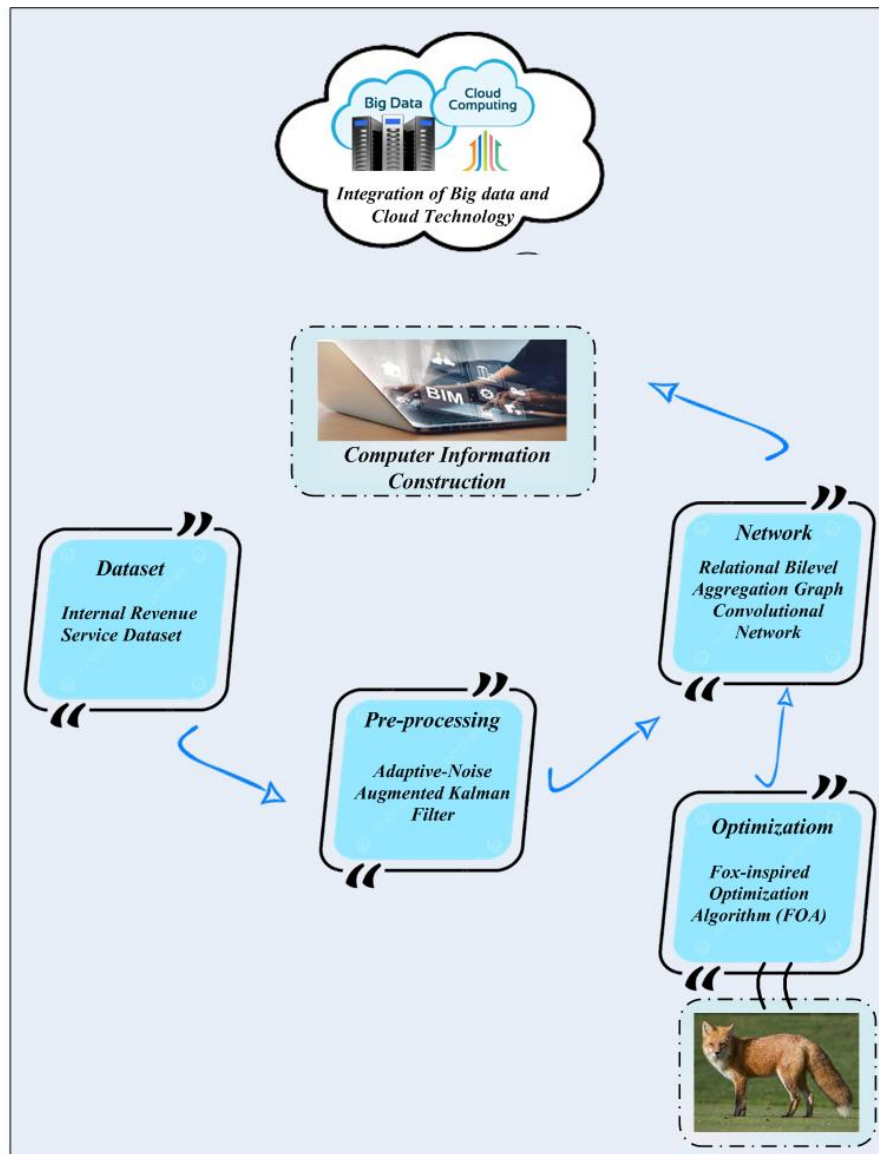


Figure1: Block Diagram for Proposed ICCBDT-CIC-RBAGCN-FOA Method

A. Data Acquisition

In this section, data are collected from Internal Revenue Service Dataset [28]. This dataset compiles IRS annual files released between 2005 and 2017. Selected income and tax components are shown in ZIP Code data, categorized by State, ZIP Code, and amount of adjusted gross income. Individual income tax returns submitted to the IRS serve as the basis for the data. To obtain a record per year, zip code, AGI group, and variable, the original data sheets which had a column for each variable, a line for each year, and an AGI group were transposed. The data for Wyoming in 2006 was eliminated since the resulting AGI classes were improperly defined and unsuitable for analysis.

B. Pre-processing Using Adaptive-Noise Augmented Kalman Filter

In this section, ANAKF [29] technique is utilized to cleaning of the collected input data. Advantages of Adaptive-Noise Augmented Kalman Filter that combine Big Data and Cloud Computing include, Optimized computer informatization construction; better real-time data processing; and increased accuracy in dynamic contexts. An important development in sensor fusion and estimating methods, especially for computer informatization building, is the ANAKF. This novel method improves the precision and dependability of estimate procedures in dynamic contexts by combining big data and cloud computing; and it given as equation (1),

$$F^0 = \frac{1}{m_0} \sqrt{\sum_{l=1}^{m_0} \left(\arg \min_{\theta} \left\| \theta x_l^0 - (x_l^0 - \hat{x}_l^0) \right\|_2 \right)^2} \tag{1}$$

Where, \hat{x}_l^0 denotes measures quantities of data; θ denotes linear function of filtering; x_l^0 denotes response vector. Because of turbulence and uncertainty, typical Kalman Filters may find it difficult to effectively estimate the state of a system in computer-intensive industries like information building, where massive volumes of data are created from many sensors and sources; and it given as equation (2),

$$F^p = \frac{1}{m_r} \sqrt{\sum_{l=1}^{m_r} \left(\arg \min_{\alpha} \left\| \alpha x_l^r - (x_l^r - \hat{x}_l^r) \right\|_2 \right)^2} \tag{2}$$

Where, \hat{x}_l^r denotes normalizing data; x_l^r selection matrix of input data; α denotes corrected data from input dataset; m_r denotes co-relation between processed data. In order to overcome this difficulty, the ANAKF dynamically modifies the data cleaning covariance matrices in accordance with the system's properties and observed data; and it given as equation (3),

$$g_h^0 = \Psi^0 P_h \tag{3}$$

Where, Ψ denotes distinguishing the measured and unmeasured data; P_h denotes generalized coordinates vector. The ANAKF can effectively process and analyse massive volumes of data in real-time by utilizing cloud computing capabilities. Decisions may be made quickly as a result, and method can effectively adjust to changing circumstances; and it given as equation (4),

$$\hat{g}_h^p = \Psi^p \Psi^{0\Phi} g_h^0 \tag{4}$$

Where, \hat{g}_h^p denotes state data vector; $\Psi^{0\Phi}$ denotes coefficient of selected data. Big data technology significantly improves the estimate process by making it easier to pull insightful information from the massive volumes of define data; and it given as equation (5),

$$F^v = \sqrt{\frac{1}{M} \sum_{j=g}^{g+M} \frac{P_j^{vv}}{\hat{v}_j^2}} \tag{5}$$

Where, P_j^{vv} denotes Adaptive-noise augmented Kalman filter estimated data; M denotes mean value of input data. By processing ANAKF method the input data's are normalized. Following preprocessing, Computer Informatization Construction receives the images.

C. Computer Informatization Construction using Relational Bilevel Aggregation Graph Convolutional Network

In this segment, RBAGCN [30] discussed to evaluate and analysis the process of computer information construction. Advantages of RBAGCN using Integration of Cloud Computing and Big Data Technology in Computer Informatization Construction include optimized resource utilization, accelerated data processing through distributed computing, and heightened predictive accuracy leveraging vast data repositories. RBAGCN, a mouthful indeed, represents a cutting-edge approach in the realm of computer informatization construction. It leverages the integration of cloud computing and big data technology to create a robust framework for analysing complex relational data; and it given as equation (6),

$$A(o) = \left(\{D_g(o), C(o)\} \right) \tag{6}$$

Where, $A(o)$ denotes structural neighbourhood of network node; $D_g(o)$ denotes graph's connected neighbourhood node; $C(o)$ denotes graph's disconnected neighbourhood node. At its core, this network architecture addresses the challenge of handling interconnected data structures, commonly encountered in modern information systems. By incorporating bilevel aggregation techniques, it can effectively capture hierarchical relationships within the data, enabling more nuanced analysis and information construction; and it given as equation (7),

$$D_g(o) = \{w | w \in V, (w, o) \in F\} \tag{7}$$

Where, V denotes utterance of information; F denotes edge of the each node; o denotes target node; w denotes features of the node. Moreover, the utilization of graph convolutional networks signifies a departure from traditional convolutional neural networks, offering enhanced capabilities in processing graph-structured data; and it given as equation (8),

$$h(w, o) = \left(1 - \frac{\arccos(\text{sim}(l_w, l_o))}{\pi} \right) (w \in M(o)) \tag{8}$$

Where, $h(w, o)$ denotes similarity if metric function; $\text{sim}(l_w, l_o)$ denotes cosine similarity of metric function; π denotes network operator. Furthermore, the integration of cloud computing and big data technologies amplifies the network's scalability and efficiency; and it given as equation (9),

$$\sigma^{(j)}(d) = (U^{(j)}d + b^{(j)}) \tag{9}$$

Where, $\sigma^{(j)}(d)$ denotes linear transformation function; $U^{(j)}d$ denotes weight matrix; $b^{(j)}$ denotes bias vector. Cloud computing provides the computational resources necessary for training and inference tasks, while big data technologies facilitate the storage and processing of massive datasets, enabling the network to handle vast amounts of information effectively; and it given as equation (10),

$$x_i = \sigma(U \cdot (i_g^o \| k_i)) \tag{10}$$

Where, x_i denotes final feature representing; k_i denotes original feature; i_g^o denotes virtual nodes. Finally, RBAGCN used to evaluate and analysis the process of computer information construction. Here, Fox--inspired Optimization Algorithm (FOA) is employed to optimize the RBAGCN. Here, FOA is employed for tuning the weight and bias parameter of RBAGCN.

D. Optimization Using Fox-inspired Optimization Algorithm

The proposed Fox-inspired Optimization Algorithm(FOA) [31] is utilized to enhance weights parameters $\sigma^{(j)}$ and x_i of proposed RBAGCN. The parameter $\sigma^{(j)}$ and x_i is implemented for increasing the accuracy and Recall. The capacity of the Fox-inspired Optimization Algorithm to effectively search complex solution spaces and its flexibility to dynamic situations, which improves convergence speed and solution correctness, are its main advantages for computer information generation. The fox is highly skilled at hunting its prey from both above and below. To find prey, the red fox initially roams the search area at random. It uses the prey's ultrasonic hearing to locate its prey. Inevitably took inspiration from this random walk and used it to provide exploratory behavior in FOX. While searching, the red fox could hear its prey.. The red fox is now in the exploitation phase after hearing the sound. Here, step by step procedure for obtaining appropriate RBAGCN values using FOA is described here. To creates a uniformly distributed population for optimizing the ideal RBAGCN parameters. The entire step method is then presented in below,

Step1: Initialization

Initial population of FOA is, initially generated by randomness. Then the initialization is derived in equation (11).

$$F = \begin{bmatrix} f_{1,1} & \dots & f_{1,j} & \dots & f_{1,p} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ f_{i,1} & \dots & f_{i,j} & \dots & f_{i,p} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ f_{N,1} & \dots & f_{N,j} & \dots & f_{N,p} \end{bmatrix}_{N \times p} \tag{11}$$

Where, f denotes the total population of fox in the tracks; p denotes the number of FOA while attacking towards its prey and N represents the distance between the prey and FOA.

Step2: Random generation

Randomly generated input parameters. Optimal fitness values were selected based on clear hyperparameter conditions.

Step 3: Fitness Function

The system's fitness is determined by the objective function. To determine the fitness function,

$$\text{Fitness Function} = \text{optimizing } [\sigma^{(j)} \text{ and } x_i] \tag{12}$$

Where, $\sigma^{(j)}$ is used for increasing the Accuracy and x_i is used for decreasing the Response Time.

Step 4: Exploration for optimizing $\sigma^{(j)}$

The emphasis of the Fox-inspired Optimization Algorithm changes from general exploration to focused refining during the exploitation phase, much like a fox hunts its prey with deliberate precision. Utilizing the information gathered during the exploration stage, the algorithm systematically targets interesting areas of the solution space to effectively and efficiently hone in on the best possible answers. This phase uses advanced optimization techniques like crossover, mutation, and adaptive strategies to hone in on and capitalize on the most promising solutions, taking its cue from the wily ways of the fox. Through astute utilization of the data acquired from exploration and dynamic modification of its search approach, the algorithm mimics the fox's capacity to capitalize on opportunities and optimize its odds of snagging prey; and it given as equation (13),

$$Y_{(it+1)} = \text{Best}X_{it} * \text{rand}(1, \text{dim ension}) * \text{Min}T * b + \sigma^{(j)} \tag{13}$$

Where, $\text{Best}X_{it}$ denotes best solution; $\text{Min}T$ denotes minimum time variable; b denotes movement of fox and. Figure 2 shows the corresponding flowchart.

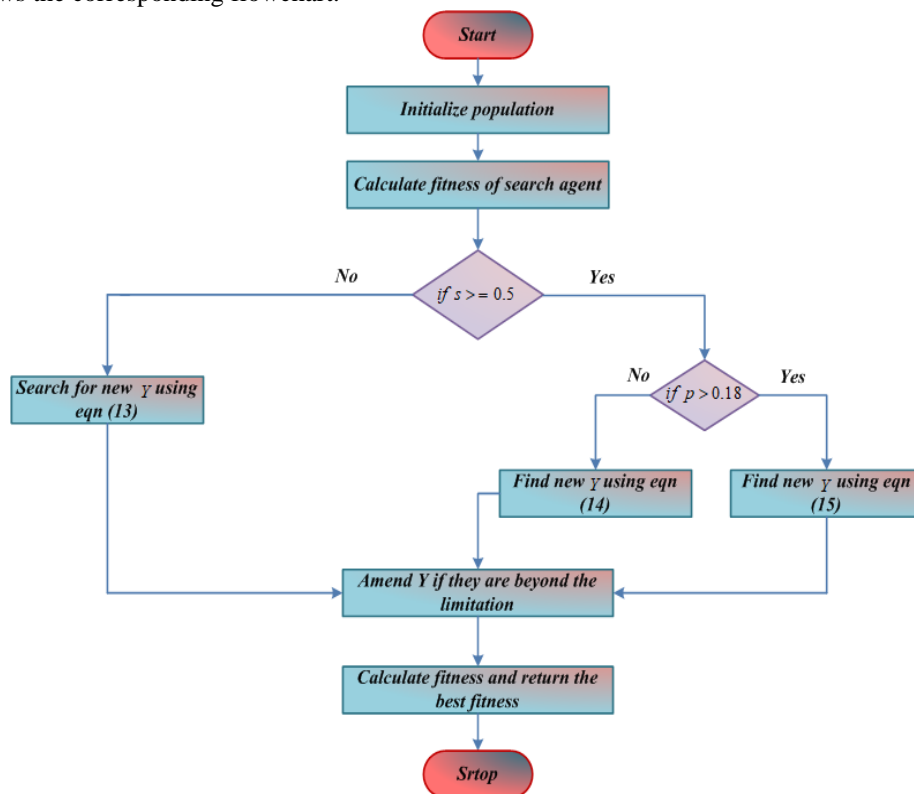


Figure2: Flow Chart of FOA for Optimizing RBAGCN

Step 5: Exploitation for optimizing x_i

The Fox-inspired Optimization Algorithm's exploration phase, which searches the solution space for optimal solutions, is reminiscent of the crafty behavior of its namesake. This phase begins with a wide range of potential solutions, inspired by the intelligence of fox behavior. This is similar to the fox's shrewd sense of exploration and adaptation in a variety of settings. By utilizing methods such as randomization and local search, the algorithm systematically explores the solution space, taking advantage of favourable areas and iteratively improving its search path. The algorithm emulates the fox's innate ability to seek prey by iteratively improving and cleverly utilizing local information. It does this by gradually focusing on potential solutions with agility and precision; and it given as equation (14),

$$Y_{(it+1)} = \text{Dist_Fox_Prey}_{it} * \text{Jump}_{it} * d_1 \cdot x_i \tag{14}$$

Where, $Prey_{it}$ denotes jumping height of prey; $Jump_{it}$ denotes jumping height of fox; d_1 denotes variable. Once the fox and prey have established their distance from one another, the red fox must relocate such that it must jump in order to capture the victim and it given as equation (14),

$$Y_{(it+1)} = Dist_Fox_Prey_{it} * x_i * Jump_{it} * d_2 \tag{15}$$

The red fox therefore approaches the worldwide ideal condition and increases its likelihood of exploiting a novel position. But the red fox leaps in the other way, toward the northeast.

Step 6: Termination Criteria

The weight parameter value of generator $\sigma^{(j)}$ and x_i RBAGCN is optimized by utilizing Fox-inspired Optimization Algorithm (FOA) and it will repeat step 3 until it obtains its halting criteria $F = F + 1$. Then ICCBDT-CIC-RBAGCN-FOA effectively construct new student management system with high Accuracy and Precision.

IV. RESULT AND DISCUSSION

Experimental results of ICCBDT-CIC-RBAGCN-FOA are discussed. The simulation is implemented in Python using Internal Revenue Service Dataset. The ICCBDT-CIC-RBAGCN-FOA model is to computer information construction using several performance metrics like Accuracy, precision, Recall, Acceleration rate, task execution rate, Response Time attained result of ICCBDT-CIC-RBAGCN-FOA method is analysed with existing techniques likes DAIMS-BDCT-RBM [21], CDP-RCR-BDA-CNN [22] and CTSC-BDCI-SVML [23] methods.

A. Performance Measures

This is a crucial step for determining the exploration of optimization algorithm. Performance measures to evaluate to access performance such as Accuracy, precision, Recall, Acceleration rate, Task execution rate, Response Time are analysed.

1) Accuracy

Accuracy describes information construction rate that are correctly constructed. The formula is derived in equation (16).

$$Accuracy = \frac{(TP + TN)}{(TP + FP + TN + FN)} \tag{16}$$

Here, TN specifies True Negative; TP specifies True Positive; FP specifies False Positive and FN specifies False Negative.

2) Precision

It estimates positive result count while computer information construction. Then the formula is derived in equation (17).

$$Precision = \frac{TP}{(TP + FP)} \tag{17}$$

3) Recall

Recall is a performance metric commonly used in convergence application. It is given in equation (18)

$$Recall = \frac{TN}{(FP + TN)} \tag{18}$$

4) Acceleration rate

The change in velocity over time is known as the acceleration rate. This allows you to get the velocity change rate in metres per second squared. Acceleration's direction and magnitude are also provided as it is a vector variable.

$$a = \frac{\Delta u}{\Delta t} \tag{19}$$

Where, Δu denotes change in velocity; Δt denotes change in time.

5) Task Execution Rate

The rate at which tasks or processes are finished in a specified amount of time is known as the task execution rate. It's frequently expressed as tasks per unit time or operations per second in computer or operations research contexts.

$$P = \frac{1}{T} \tag{20}$$

Where, P denotes rate of work completion (tasks per unit time); T denotes time needed to finish a single task.

6) *Response Time*

Divide the entire time (in minutes) you spent responding to tickets during a week by the total number of tickets to determine the average response time. Your average response time on a series basis is the outcome.

$$PT = RT + ST + VT \tag{21}$$

Where, RT denotes processing time; ST denotes queuing time; VT waiting time.

B. *Performance analysis*

Figure 3 to 8 shows simulation result of ICCBDT-CIC-RBAGCN-FOA method. The performance metrics are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods.

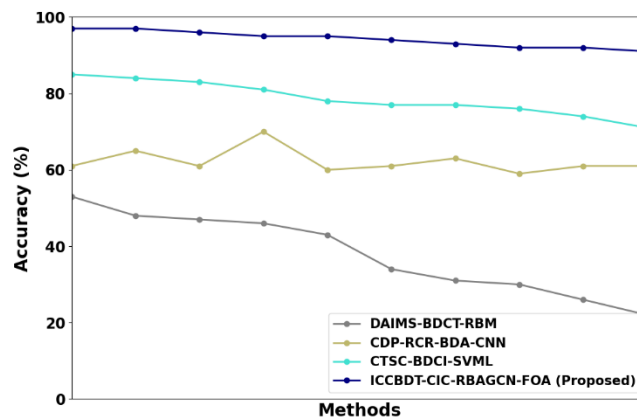


Figure 3: Performance analyses of Accuracy

Figure 3 depicts accuracy analyses. In computer information constructing, the above an accuracy graph primarily shows how effectively an algorithm performs across a range of conditions or inputs. It provides a visual representation of how well the system exist results in relation to known or expected values. Usually, the y-axis indicates the corresponding accuracy levels, and the x-axis shows various inputs. The proposed ICCBDT-CIC-RBAGCN-FOA attains 18.57%, 22.15% and 31.10% higher Accuracy for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods respectively.

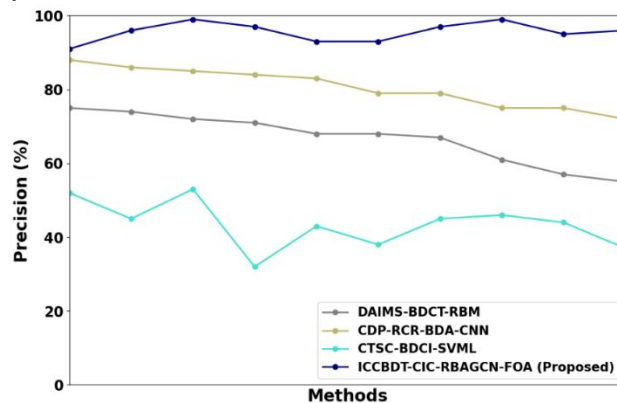


Figure 4: Performance analyses of Precision

Figure 4 depicts Precision analyses. In computer information construction, the above precision graph shows how accurate a proposed method is in different datasets. It shows how reliably the system yields accurate results or distinguishes pertinent data from irrelevant data. Generally, the precision levels attained are displayed on the y-axis, and various methods are displayed on the x-axis. The proposed ICCBDT-CIC-RBAGCN-FOA attains

19.59%, 23.12% and 32.60% higher Precision for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods respectively.

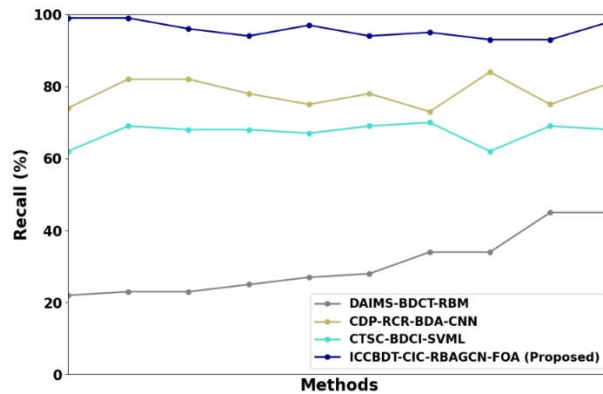


Figure 5: Performance analyses of Recall

Figure 5 depicts Recall analyses. Plotting the above recall graph's rate versus the total number of data retrieved creates a recall graph, which is used in computer information construction to illustrate the effectiveness of a retrieval system. It provides a visual representation of how successfully the system can extract pertinent information of a dataset. As more items are retrieved, the recall rate tends to increase on the graph; the slope of the line indicates how well the system finds the relevant data. The proposed ICCBDT-CIC-RBAGCN-FOA attains 19.57%, 25.15% and 31.60% higher Recall for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods correspondingly.

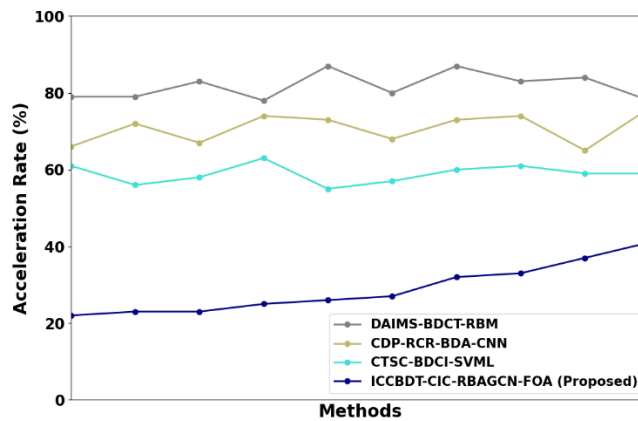


Figure 6: Performance analyses of Acceleration rate

Figure 6 depicts accuracy Acceleration rate. The above acceleration rate graph, which is created using computer data, shows how acceleration changes over time for a specific suggested model. Usually, it plots various methods on the x-axis versus acceleration on the y-axis. This acceleration rate graph offers insights on the dynamics and performance of input data collection. The proposed ICCBDT-CIC-RBAGCN-FOA attains 18.59%, 23.95% and 31.20% lower Acceleration rate for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods correspondingly.

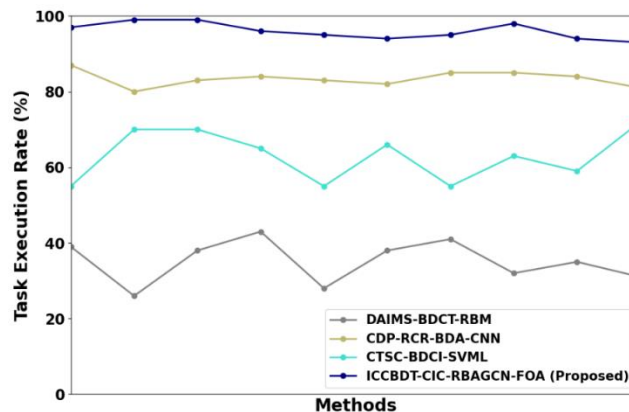


Figure 7: Performance analyses of Task execution rate

Figure 7 depicts Task execution rate analysis. During computer information construction, the above task execution rate graph shows the throughput and efficiency of task completion over time. It usually shows the speed at which a system completes tasks. This graph offers insightful information on performance trends. The proposed ICCBDT-CIC-RBAGCN-FOA attains 18.57%, 23.15% and 31.60% higher Task execution rate for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods respectively.

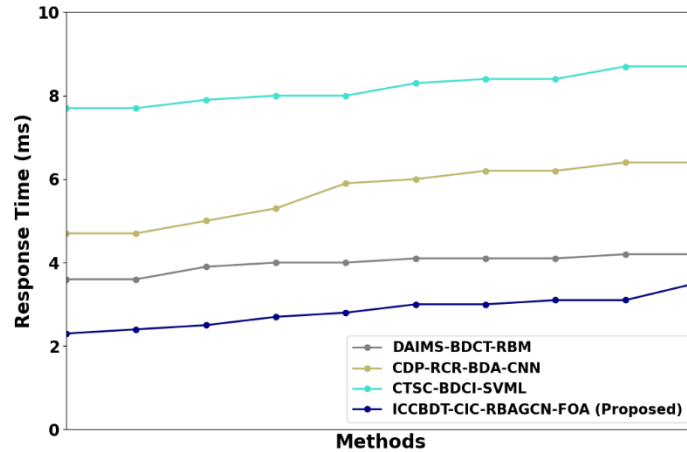


Figure 8: Performance analyses of Response Time

Figure 8 depicts Response Time analysis. In computer information construction, the above response time graphs show how long it takes a system to react to a request over time. The above graph provides insights into efficiency, making them vital tools for evaluating system performance. Response time graphs, which typically show various methods on the x-axis and response duration on the y-axis, enable users to monitor variations inefficiencies in the system. The proposed ICCBDT-CIC-RBAGCN-FOA attains 19.57%, 22.15% and 31.90% lower Response Time for computer information construction which are analysed with existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML methods respectively.

C. Discussion

A novel ICCBDT-CIC-RBAGCN-FOA model to constructs computer information based on big data and cloud computing technology. The ICCBDT-CIC-RBAGCN-FOA method involves encompasses ANAKF depend Internal Revenue Service Dataset pre-processing; then, RBAGCN model utilized for performing information construction which construct computer information based big data and cloud computing technology. The computer information construction of Internal Revenue Service Dataset, average highest results of method were compared to average outcomes given in existing techniques likes existing DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML. These existing are high expensive than comparing to the proposed method. However, the proposed technique employs faster RBAGCN in conjunction by FOA, resultant in further efficient collection of data, enhanced ability to deal by method over-fitting problem. Therefore, the comparative methods are expensive than the proposed technique. As a result, the proposed technique construct computer information using optimizes neural network more effectively and efficiently.

5. CONCLUSION

In this section, big data and cloud computing technology based computer information construction (ICCBDT-CIC-RBAGCN-FOA) is successfully executed. The simulation is implemented in Python. According to the experimental results ICCBDT-CIC-RBAGCN-FOA performed better when used with the Co-training technique than when used separately regards Response Time and Task execution rate. The performance of ICCBDT-CIC-RBAGCN-FOA approach attains 19.57%, 22.15% and 31.90% lower Response Time, 18.57%, 23.15% and 31.60% high Task execution rate, 18.59%, 23.95% and 31.20% lower Acceleration rate when analysed with existing methods like DAIMS-BDCT-RBM, CDP-RCR-BDA-CNN and CTSC-BDCI-SVML respectively. As part of our future research, intend to examine the difficulties associated with putting data collection into practice in densely populated nations and show how big data and cloud computing can be used effectively in areas like data processing, collection, and analysis to solve pressing issues and unresolved issues with infrastructures for smart cities.

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