

A Comprehensive Analysis of Proprietary and Open Source Data Mining Tools

Sonia Rani Chowdhary, Vikash

Computer Application, Post Graduate Government College, Chandigarh, Chandigarh, India

ABSTRACT

The Powerful software tools and techniques required for the development of data mining applications. With the rapid development of technologies and business interest in using electronics and latest technologies plays important role in improvement of data mining field. Data mining access the meaningful and efficient information available in worldwide which is helps in decision making. This paper described the (a) various tools and techniques used by data mining applications. (b) compared features and limitations both in Proprietary and open sources data mining tools. (c) technical analysis of proprietary and open source data mining tools. On the basis of well-designed User interface, short time analysis, statistical and mathematical analysis user can select the best tool as per their requirements. Analysis of these tools makes easy to select appropriate tool.

Keywords : Data Mining Tools, Open Source Tools, Proprietary tools, Technical Specifications

I. INTRODUCTION

Today with the development of technologies the needs of data become more important in everyday life. The size of data increases day by day. Amount of information double every 20 months according to estimations [1]. Data mining is process of discovering useful pattern among large amount of data such as Big data and transformed into more informative form. It is a core step of knowledge discovery. Global IP traffic is expected to reach more than 190 exabytes per month in 2020 according to Cisco Forecast [2] and in 2025 total amount of data expected to reach over 160 zettabytes [3]. Worldwide market for Business Intelligence (BI) accounted over 17 Billion USD in 2016 according to Gartner and 2021 it expected to reach more than 26 Billion USD, where modern BI and analytics expands more rapidly than other fields [4].

Data mining is not only important in BI but also in technologies and life science sectors. It is strongly associated with data science including modifications and classifications of data by using statistical and mathematical concepts. Data mining is also important in public opinion and marketing research [5] [6].

The purpose of this paper to not only comprehensive comparison of all proprietary and open source tools but also analysis each and every tool, which helps to select best tool for specific field or area.

II. LITERATURE'S REVIEWS

A Comparative Study of Famous Classification Techniques and Data Mining Tools on the basis of specific parameters. Paul Y, Kumar N described various classifications techniques and data mining tools along with their advantages and disadvantages. KNN (K- Nearest Neighbor) classifier is good choice for classifications problem where the data set size is small and noise free. SVM (Support Vector Machine) are frequent and widely used classifier. Decision tree and Bayesian network discussed. They represent many combined methods but still not clear which one is best to full fill the requirements [7]. Myint Myint Than represents a paper called "A Comparative Study of predicting teaching score by using classification algorithms" [8] this paper respectively expounds comparison between Naïve Bayesian, J48 and Random Forest Decision Tree classifiers are shown on the frequent item set is developed for data analysis. The data set analyze for better prediction. According to this paper result factors, the performance of J48 Decision tree classifier is more accurate and more precision than Naïve Bayes and Random Forest classifiers. J48 Decision tree algorithm is best for educational manager for future decision making in teacher selection by using its resulting data. Feng, Junshuai, introduced a paper called "Predicting Students' Academic Performance with Decision Tree and Neural Network" [9]. This paper explained EDM (Educational Data Mining) and Neural Networks in data mining applications. EDM and Neural Networks related to educational background and can analyze and resolving educational data problems. Both are utilized in widespread and successful data mining applications. The aim of this paper to explore the methodology such as decision tree, classifiers and neural network to predict student performance in the context of EDM. Sibel Barin Özkan, Sultan Muhammed et al. [10] explain in this paper many commercial and open source program to implement Data Mining application. This paper introduced open source data mining programs WEKA, Orange, KNIME. Aim of this study to determine the difference between these three data mining programs with Naïve Bayes classification algorithms by taking Bay iris, breast cancer, wine, monk, balance es data set from UCI Machine Learning Repository database. Hansheng Lei, Mahmoud Quweider et al. introduced paper "Mining Survey Data" [11] in this represents the strategies in mining survey data using computational methods. A Novel method for data preparation and dependent pattern mining is presented. It introduced DP (Dependent Pattern) algorithm that is not only mines compact sets of patterns, but also discovers meaningful patterns.

III. OVERVIEW OF PROPRIETARY DATA MINING TOOLS

Sisense is best suited for business intelligence software. Non-technical persons can easily use it by its drag and drop facility. SSDT (SQL Server Data Tools) has all phases of database. It is a declarative model and has visual tool for development like intelligence, code navigation tools and programming support via c#, visual basic etc. Oracle data mining is based on relational database. It has algorithms for data classifications, prediction, regression and specialized analytics which helps to make better prediction. IBM Cognos helps to meet organization needs like Cognos connections, query studio, report studio, analysis studio, event studio and workspace advance. IBM SPSS Modeler produced by SPSS and then acquired by IBM. It has visual interface and no need of programming knowledge to interact with it. SAS data mining is analysis and data management SAS institute develop Statistical Analysis System (SAS). It helps to take timely decision. Tera data mining follow no sharing structure it means it has their own memory and processing ability. Board gather data from heterogeneous resources and analyze it. Dundas BI is a quick insight tool. It has Gap free protection of document feature which provides good accessibility across many devices [12].

IV. OVERVIEW OF OPEN SOURCE DATA MINING TOOLS

Rapid Miner is one of the best predictive analysis. It is based on client/server model. Orange is components based and data visualization tool. It helps users to take smart decision. Weka is also known as Waikato environment. SQL database access features also available in Weka. KNIME (Konstanz Information Miner) is basically used for pharmaceutical research but also used in other areas like business intelligence and financial data analysis. Apache Mahout has clustering, classification and collaboration filtering features. Rattle tool generate duplicate code for any activity happen in GUI by using inbuilt language code tab. Data Melt is known as DMelt. This tool is basically for engineers, scientist and students. KEEL (Knowledge Extraction based on Evolutionary Learning) used to discover large number of different knowledge tasks. R(Revolution) data mining is very useful for complicated analysis without knowing the depth of computer system [13].

V. COMPARATIVE ANALYSIS OF PROPRIETARY AND OPEN SOURCE DATA MINING TOOLS

The best available proprietary and open source data mining tools were chosen and analytical study was made by taking into account technical specifications and features

Sno.	Tool Name	Technologies/Algorithms	Language	Features	Limitations
1.	Sisense	In-Chip analytics (In	• ETL	Data Size limitations removed	• Size of Sisense application is
	[19]	Memory Technology)	Python	Unifying Data	very heavy
		Prism 10X	• R	Advanced analytics	 Sisense's dashboards only
		 Crowd Accelerated BI 		• Wide range of widgets such as	interact on the web
		 Sisense Pulse 		gauges, charts and graphs	 Navigation and filtering on
				Complex business queries without	mobile platform should be
				programming or SQL writing	improved.
					• Most of researchers are not
					familiar with it.s
2	SSDT	SOI Server database	• 501	Can create offline Database	Required knowledge of
2.	(SOL	technology	• SQL	Analysis Services (AS) data	database system
	Server Data	• Azure SOL databases		models	 DAC (distributed client-
	Tool) [20]	 Analysis Services (AS) 		 Integration Services (IS) packages 	server architecture)
		data models		 Paporting Services (PS) reports 	framework compatibility
		 Integration Services (IS) 		• Reporting Services (RS) reports	issues
		nackages			
		Reporting Services (RS)			
3.	Oracle Data	Naive Bayes	PL/SQL	Association Model Aggregation	Required knowledge of
	Mining	Support Vector Machine	 Java APIs 	Metrics	database system.
	[21]	Decision Tree	• SOL	Support for Explicit Semantic	Navigation system should be
		 Enhanced K-Means 		Analysis Algorithm	improved
		Orthogonal Partitioning		Support for Partitioned Models	
		Clustering			
		Expectation			
		Maximization			
		Apriori			
		Non-negative Matrix			
		Factorization			
		 Principal Components 			
		Analysis (PCA)			
		 Singular Vector 			
		Decomposition			
4.	IBM	Correlation	• Java	 Visualize organizational 	• Its file become very large in
	Cognos	 decision-tree algorithms 		performance	size
	[22]	 predictive algorithms 		Protect data	 Limited data size used
		 machine learning 		Share critical insights easily	
		 natural language 		Automated data preparations	•
		processing			
5	IBM SPSS	decision trees	Python	Support for many data sources	Text Analytics node
2.	Modeler	 neural networks and 	- i yulon	Easy model deployment	restrictions
	[23]	regression models		 Powerful graphics engine 	 job will fail if the firewall
		Generative adversarial		Automatic data preparation	blocks the connection
		networks (GANs)			between the two servers.

Table-1: Technical Overview proprietary data mining tools

					• Navigation system should be improved
6.	SAS Data Mining [24]	 cutting-edge algorithms Clustering Anomaly detection Association rule learning Principal component analysis Affinity grouping Regression Neural networks Decision trees Support vector machines 	CPythonR	 Easy-to-use GUI and batch processing Model comparisons, reporting and management. Sophisticated data preparation, summarization and exploration. Automated scoring 	 Lack of graphic representation Difficult Text Mining Difficult than R Costly
7.	Tera Data [25]	 Cutting-edge technology Machine Learning Algorithms 	 SQL R Python with Scala Go JavaScript 	 Unlimited Parallelism Shared Nothing Architecture Linear Scalability Robust Utilities 	 Multiload utility limitation Lack of graphic representation
8.	Board [26]	 cutting-edge technology in-memory technology called HBMP (Hybrid Bitwise Memory Pattern) combination of HTML5 technology with the "Toolkit" 	• Python	 Multi-language Single logical view of corporate data Data Connectivity Granular security Server Clustering 	Limited reporting capabilityBig sizeData limitation
9.	Dundas BI [27]	 HTML5 web technology server-based technology Microsoft Technology Integration 	 ETL Python R .Net 	 Natural Language User Input Table Relationships Multi-Tenancy Improvements Heat Map Chart 	 Not flexible with 3D charts Require large server RAM for memory processing Does not support text analytics
10.	Rapid Miner [28]	 machine learning algorithms classification and clustering algorithms k-Means Clustering 	Language Independent	 More than 20 new functions for analysis and data handling, including multiple new aggregation functions File operators to operate directly from Rapid Miner A macro viewer that shows macros and their values in real time during process execution Intuitive GUI 	Requires best knowledge of database management.
11.	Orange [29]	 machine learning algorithms classification and regression algorithms K-nearest neighbor (KNN) Random Forest Neural Network Naïve Bayes and CN2 Rule Inducer 	 Python C++ C 	 Visual Programming, Visualization, Interaction and Data Analytics Large toolbox, Scripting interface Extendable Documentation 	 Limited reporting capability Very big installation.
12.	Weka [30]	 machine learning algorithms classification/regression Clustering searching Algorithms 	• Java	 Forty-nine data preprocessing tools, seventy-six classification/regression algorithms, eight clustering algorithms, fifteen attribute/subset evaluators, ten search algorithms for feature selection. three algorithms for finding association rules Good documentations- a whole textbook being devoted to it. 	 Weak classical statics Weak CSV reader.
13.	KNIME [31]	 machine learning algorithms Tree Ensemble Decision Tree 	• Java	 Scalability, Intuitive user interface, High extensibility well-defined API for plugin extensions 	Poor parameter optimizationLimited error measurements

		Random Forest, and Naïve Bayes learner and predictor nodes		 sophisticated data handling, intelligent automatic caching of data, Data visualization Import/export of workflows, Parallel execution on multi-core systems Command line version for "headless", "batch executions", Hilting, 	no wrapper methods for descriptor selection
14.	Apache Mahout[32]	 Clustering, classification batch based collaborative filtering, 	• Java • Scala	 Extensive programming environment Pre-made algorithms Math experimentation environment 	 Limited Algorithms User interface should be improved
15.	Rattle[33]	 Decision trees, boosting random forests support vector machines generalised linear models neural networks 	• R	 Graphical user interface Data Visualization Statistical power of R 	 Limited data sources used Generate duplicate code for any activity in GUI
16.	Data Melt [34]	 Mathematical and statistical algorithms Linear regression, curve fitting cluster analysis, neural networks fuzzy algorithms analytic calculations and interactive visualisations using 2D/3D plots histograms 	 Python Ruby Groovy java 	 Numeric computation, statistics, analysis of large data volumes ("big data") and scientific visualization. computational platform creates high-quality vector-graphics images 	 Data limitations Limited algorithms and formulas used
17.	KEEL [35]	 Evolutionary algorithms extraction algorithms Computational Intelligence based learning algorithms, 	• Java	 Classification Discovery Cluster Discovery Regression Discovery Association Discovery Data Visualization Discovery Visualization a user-friendly graphical interface evolutionary learning 	Limited Algorithms
18.	R[36]	Machine learning algorithms, Naïve Bayes algorithm, confusion matrix	 C Fortran R 	 Data Exploration Outlier detection Clustering Text Mining Time Series Analysis Social Network Analysis Parallel Computing, Graphics Visualization of geo spatial data Web Application Big data Data and error handling 	 Requires Knowledge of array language Less specialized for Data Mining.

Table-1 represents the technical overview of the proprietary and open source data mining tools which includes name of tool and description of technologies and algorithms, language.

VI. ANALYSIS

This paper examined some best proprietary and open source data mining tools. Proprietary tools are marketed and protected by registered trade and open source tools are freely accessed by the user. Every tool has their own features and limitations. Sisense is a business intelligence tool. It processes data small to large organizations. Non-technical user easily operates it. But very less number of users used it. SSDT (SQL Server Data tool) is declarative model that is used for expand the database phases. It has capability to maintain, debug and refactor database. Basic database management knowledge required for SSDT. Oracle data mining is also based on database. It is used for better prediction, identify cross-selling. User can easily interact with it just because of it has drag and drop feature. IBM Cognos is Intelligence suite for reporting and data analysis. IBM SPSS Modeler is data mining and text analysis tool for built predictive model. It has a visual interface that allow user to work with data mining. SAS Data Mining is for statistical analysis. It helps user to make timely decision. Tera Data is used for business analysis. Its working on share nothing architecture its server node has their own memory and processing ability. Board is best tool for companies looking for improvement in decision making. Its interface is more attractive and comprehensive interface as compare to all BI software. Dundas BI is reporting and analytics tool. It can create attractive tables and charts. It has gap free protection of data.

Rapid Miner and Orange tools are considered appropriate for advance users with knowledge of science because additional skills are required. There are very limited visualizations support present in these tools. Weka is having many inbuilt features and there is no programming knowledge required. It is auto parameterized and have very good documentations. KNIME is highly recommendable for the users who are highly skilled. It is very robust with inbuilt features and can obtain third party library functions. Apache Mahout is for machine learning algorithms. It can perform mathematical functions. Rattle is very user friendly it has well developed UI. Data Melt has interactive framework for data analysis. It is a multiprogramming utility. KEEL is software tool to access evolutionary algorithms for data mining

problems. It is restricted by the number of algorithms. R is well-designed statistical data mining tool. It is less specialized for data mining.

VII. CONCLUSION

There are many proprietary and open source data mining tools available. In this paper different types of data mining tools are explained. They all have their features and limitations. Most of the tools provides nice graphical user interface, focus on usability, interactivity and security of data. Data visualization is great feature of software. Statistical and mathematical functions are key features of some tools. User can use the software as per their requirement. There are many proprietary and open source tools available. According to user needs and requirements they can used them. Increase of data mining tools is still challenges for select best tool for specific use. For future work makes data mining tools more efficient, interactive, user friendly, consume less space and complete work in short time.

VIII. REFERENCES

- Kalpana Rangra, Dr. K. L. Bansal "Comparative Study of Data Mining Tools", International Journal of Advanced Research in Computer Science and Software Engineering, (June 2014); Volume 4, Issue 6
- [2]. San Jose, CA Cisco "The zettabyte era: trends and analysis." Cisco., Visual Networking Index. (2017).
- [3]. Reinsel, D., Gantz, J., & Rydning, J. "Data age 2025: The evolution of data to life-critical". Framingham, MA: International Data Corporation, (2017).
- [4]. Cheung, M., Dharmasthira, Y., Eschinger, C., Anderson, R. P., Granetto, B. F., Adams, A., ... Roth, C. "Forecast: Enterprise software markets, worldwide", 2013–2020. Stamford, CT: Gartner RAS Core Research Notes Gartner, (2016).
- [5]. Ravi, K., & Ravi "A survey on opinion mining and sentiment analysis: Tasks, approaches and

applications". Knowledge-Based Systems, (2015);89, 14–46.

- [6]. Scott, J." Social network analysis. Thousand Oaks", CA: SAGE,(2017).
- [7]. Paul Y., Kumar N. (2020)" A Comparative Study of Famous Classification Techniques and Data Mining Tools." In: Singh P., Kar A., Singh Y., Kolekar M., Tanwar S. (eds) Proceedings of ICRIC (2019);Lecture Notes in Electrical Engineering, vol 597. Springer, Cham
- [8]. Myint Myint Than" A COMPARATIVE STUDY OF PREDICTING TEACHING SCORE BY USING CLASSIFICATION ALGORITHMS" Lecturer, University of Computer Studies, Kalay, Burma , (2019); IJARIIE-ISSN(O)-2395-4396, Vol-5 Issue-4
- [9]. Feng, Junshuai, "Predicting Students' Academic Performance with Decision Tree and Neural Network" (2019); Electronic Theses and Dissertations. 6301. Spring Term 2019
- [10]. Sibel Barin Özkan, Sultan Muhammed Fatih Apaydin, Yasin Özkan, Irem Düzdar" Comparison of Open Source Data Mining Tools: Naive Bayes Algorithm", Scientific Meeting on Electrical-Electronics & Biomedical Engineering and Computer Science (EBBT), IEEE (2019).
- [11]. Hansheng Lei , Mahmoud Quweider et al. "Mining Survey Data" 2nd International Conference on Data Intelligence and Security (ICDIS), IEEE (2019)
- [12]. https://www.softwaretestinghelp.com/data-miningtools/
- [13]. Andreas Bartschat ,Markus Reischl, Ralf Mikut. " Data mining tools", Institute for Automation and Applied Informatics, Karlsruhe Institute of Technology, EggensteinLeopoldshafen, Germany, (4 January 2019)
- [14]. Jovic, K. Brkic and N. Bogunovic. "An overview of free software tools for general data mining", Faculty of Electrical Engineering and Computing, University of Zagreb / Department of Electronics, Microelectronics, Computer and Intelligent Systems, Unska 3, 10 000 Zagreb, Croatia, (May 2014); 26-30
- [15]. Mrs. G. SangeethaLakshmi1, Ms. M. Jayashree2, " Comparative Analysis of Various Tools for Data Mining and Big Data Mining", International Research Journal of Engineering and Technology

(IRJET), (Apr 2019); e-ISSN: 2395-0056, p-ISSN: 2395-0072, Volume: 06 Issue: 04

- [16]. Aderibigbe Israel Adekitan*, Jeremiah Abolade and Olamilekan Shobayo "Data mining approach for predicting the daily Internet data traffic of a smart university", Adekitan et al. J Big Data (2019); 6:11
- [17]. Graham J Williams "Rattle: A Data Mining GUI for R" The Research Journal · (December 2009)
- [18]. Maqsud S. Kukasvadiya, Dr.Nidhi H.
 Divecha, "Analysis of Data Using Data Mining tool",
 IJEDR (2017); Volume 5, Issue 2 | ISSN: 2321-9939.
- [19]. https://www.sisense.com
- [20]. https://docs.microsoft.com/en-us/sql/ssdt/sql-serverdata-tools?view=sql-server-ver15
- [21]. https://www.oracle.com/database/technologies/adva nced-analytics/odm.html
- $[22]. \ https://www.ibm.com/products/software$
- [23]. https://www.ibm.com/products/spss-modeler
- [24]. https://www.sas.com/en_in/software/enterpriseminer.html
- [25]. www.teradata.com
- [26]. www.board.com
- [27]. https://www.dundas.com/dundas-bi/
- [28]. https://rapidminer.com/
- [29]. https://orange.biolab.si/
- [30]. www.cs.waikato.ac.nz/~ml/weka
- [31]. www.knime.org
- [32]. mahout.apache.org
- [33]. rattle.togaware.com
- [34]. https://jwork.org/dmelt/
- [35]. https://sci2s.ugr.es/keel/download.php
- [36]. https://www.r-project.org

Cite this article as :

Sonia Rani Chowdhary, Vikash, "A Comprehensive Analysis of Proprietary and Open Source Data Mining Tools", International Journal of Scientific Research in Computer Science, Engineering and Information Technology (IJSRCSEIT), ISSN : 2456-3307, Volume 6 Issue 2, pp. 414-420, March-April 2020. Available at doi : https://doi.org/10.32628/CSEIT206210 Journal URL : http://ijsrcseit.com/CSEIT206210