



Knowledge Acquisition and Privacy Preserving in Cloud using Simultaneous Diagonalization Algorithm

Dr. Neelu Khare, Kumaran U, M. Mohan Vamsi
SITE, VIT University, Vellore, Tamilnadu,
India

Abstract— In the recent times the WWW(World Wide Web) has transformed from static collection of HTML data to a dynamic system that offered a platform for cloud technologies and distributed information systems. This paper describes how data mining is used in cloud computing while preserving its privacy. Data Mining is used for extracting potentially useful information from raw data. The integration of data mining techniques into normal day-to-day activities has become common place. Every day people are confronted with targeted advertising, and data mining techniques help businesses to become more efficient by reducing costs. Data mining techniques and applications are very much needed in the cloud computing paradigm. The implementation of data mining techniques through Cloud computing will allow the users to retrieve meaningful information from virtually integrated data warehouse that reduces the costs of infrastructure and storage.

Keywords—Privacy Preserving, Knowledge Acquisition, Data Mining, Cloud Computing, Simultaneous Diagonalization

I. INTRODUCTION

Internet is a vital tool in both professional and personal life. Most of the businesses are increasingly conducted over internet. One of the most revolutionary concepts of internet over recent years is cloud computing.

Cloud computing involves computing resources- hardware and software - that are delivered as a service over internet.

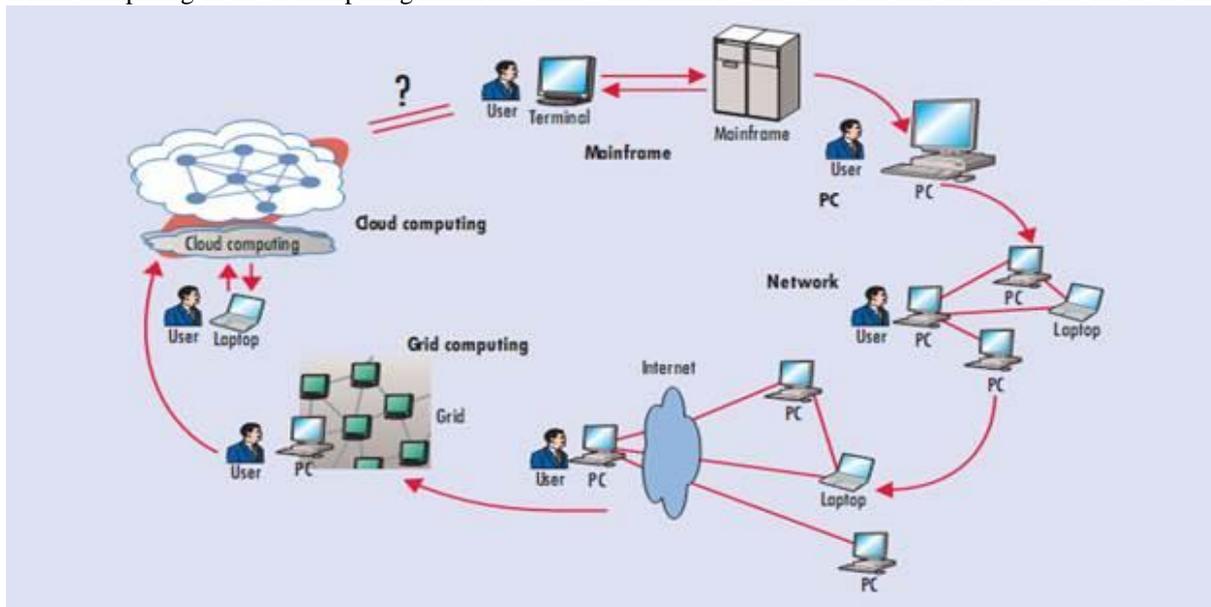


Figure 1 : Cloud Paradigm Shift of the last half century.

II. LITERATURE SURVEY

Top Cloud Computing Companies and its key features [1]

Cloud Name	Key Features
Sun Micro systems Cloud	More available application than any other open Operating Stsem
IBM Cloud Engine	Integrated power management helps us plan, predict, monitor and actively manage power consumption of BladeCenter server.
Amazon EC2	Designed to make webscale computing easier for developers.
Google App Engine	No limit to the free trial period if you do not exceed the quota allotted.

Microsoft Azure	Currently offering a “development accelerator” discount plan. 15-30 % discount off consumption charges for first 6 months.
AT&T Synaptic Hosting	Use fully on-demand infrastructure or combine it with dedicated components to meet specialized requirements.
GoGrid Cloud Computing	Free load balancing and free 24/7 support.
Salesforce	Offers cloud solutions for automation, customer service and platform, respectively. Transparency through real-time information on system performance and security at trust.salesforce.com

Data Mining Techniques in Cloud [2]

Technique Name	Applicability
Clustering	Useful for exploring data and finding natural groupings. Members of a cluster are more like each other than they are like members of a different cluster. Common examples include finding new customer segments and life sciences discovery.
Classification	Most commonly used technique for predicting a specific outcome such as response / noresponse, high / medium / low value customer, likely to buy / not buy.
Association	Find rules associated with frequently cooccurring items, used for market basket analysis, cross-sell, root cause analysis. Useful for product bundling, in store placement, and defect analysis.
Regression	Technique for predicting a continuous numerical outcome such a customer lifetime value, house value, process yield rates.
Attribute Importance	Ranks attributes according to strength of relationship with target attribute. Use cases include finding factors most associated with customers who respond to an offer, factors most associated with healthy patients.
Anomaly Detection	Identifies unusual or suspicious cases based on deviation from the norm. Common examples include health care fraud, expense report fraud, and tax compliance.
Feature Extraction	Produces new attributes as linear combination of existing attributes. Applicable for text data, latent semantic analysis, data compression, data decomposition and projection, and pattern recognition.

III. PROBLEM STATEMENT

Feature Extraction in Knowledge Discovery System

Generally, feature extraction for classification can be seen as a search process among all possible transformations of the feature set for the best one, which preserves class separability as much as possible in the space with the lowest possible dimensionality.[3].

In other words we are interested in finding a projection w :

$$Y = w^T \cdot X$$

where y is a $p' \times 1$ transformed data point (presented using p' features), w is a $p \times p'$ transformation matrix, and x is a $p \times 1$ original data point (presented using p features).

The conventional PCA gives high weights to features with higher variabilities irrespective of whether they are useful for classification or not. This may give rise to the situation where the chosen principal component corresponds to the attribute with the highest variability but having no discriminating power.[4].

A usual approach to overcome the above problem is to use some class separability criterion, e.g. the criteria defined in Fisher linear discriminant analysis and based on the family of functions of scatter matrices:

$$J(w) = \frac{w^T S_B w}{w^T S_W w}$$

where S_B is the between-class covariance matrix that shows the scatter of the expected vectors around the mixture mean, and S_W is the within-class covariance, that shows the scatter of samples around their respective class expected vectors.

IV. PROPOSED SOLUTION

Managing Feature Extraction and Privacy Preservation

Currently, as far as we know, there is no feature extraction technique that would be the best for all data sets in the classification task. Thus the adaptive selection of the most suitable feature extraction technique for a given data set needs further research. Currently, there does not exist canonical knowledge, a perfect mathematical model, or any relevant tool to select the best extraction technique. Instead, a volume of accumulated empirical findings, some trends, and some dependencies have been discovered.

We consider a possibility to take benefit of the discovered knowledge by developing a decision support system based on the methodology of expert system design in order to help to manage the data mining process. The main goal of the system is to recommend the best-suited feature extraction method and a classifier for a given data set. Achieving this goal produces a great benefit because it might be possible to reach the performance of the wrapper type approach by using the filter approach. In the wrapper type approach the interaction between the feature selection process and the construction of the classification model is applied and the parameter tuning for every stage and for every method is needed.[5]. In the filter approach the evaluation process is independent from the learning algorithm and the methods, and their parameters' selection process is performed according to a certain set of criteria in advance. However, the additional goal of the prediction of model's output performance requires also further consideration.

V. EXPERIMENTS AND RESULTS

Generally, the knowledge base is a dynamic part of the decision support system that can be supplemented and updated through the knowledge acquisition and knowledge refinement processes. Potential contribution of knowledge to be included into the KB might be found discovering a number of criteria from the experiments conducted on artificially generated data sets with pre-defined characteristics.

The results of experiments can be examined looking at the dependencies between the characteristics of a data set in general and the characteristics of every local partition of the instance space in particular. Further, the type and parameters of the feature extraction approach best suited for the data set will help to define a set of criteria that can be applied for the generation of rules of KB.

The results of our preliminary experiments support that approach. The artificially generated data sets were manipulated by changing the amount of irrelevant attributes, the level of noise in the relevant attributes, the ratio of correlation among the attributes, and the normality of the distributions of classes. In the experiments, supervised feature extraction (both the parametric and non parametric approaches) performed better than the conventional PCA when noise was introduced to the data sets. [6].

The similar trend was found with the situation when artificial data sets contained missing values. The finding was supported by the results of experiments on the LED17, Monk-3 and Voting UCI data sets (Table 1) that are known as ones that contain irrelevant attributes, noise in the attributes and a plenty of missing values.

Thus, this criterion can be included in the KB to be used to give preference to supervised methods when there exist noise or missing values in a data set. Nonparametric feature extraction essentially outperforms the parametric approach on the data sets, which include significant non-normal class distributions and are not easy to learn.

This initial knowledge about the nature of the parametric and nonparametric approaches and the results on artificial data sets were supported by the results of experiments on Monk-1 and Monk-2 UCI data sets.

Dataset	PCA	Par	NPar	Plain
LED17	.395	.493	.467	.378
MONK-1	.767	.687	.972	.758
MONK-2	.717	.654	.962	.504
MONK-3	.939	.990	.990	.843
Voting	.923	.949	.946	.921

VI. CONCLUSION

We considered the goals of such a system, the basic ideas that define its structure and methodology of knowledge acquisition ,privacy preservation and validation. The Knowledge Base is the basis for the intelligence of the decision support system. That is why we recognised the problem of discovering rules from the experiments of an artificially generated data set with known predefined simple, statistical, and information theoretic measures, and validation of those rules on benchmark data sets as a prior research focus in this area. It should be noticed that the proposed approach has a serious limitation. Namely the drawbacks can be expressed in the terms of fragmentariness and incoherence (disconnectedness) of the components of knowledge to be produced.

VII. FUTURE WORK

We do not claim the completeness of our decision support system. Otherwise, certain constrains and assumptions to the domain area were considered, and the limited sets of feature extraction methods, classifiers and data set characteristics were considered in order to guarantee the desired level of confidence in the system when solving a bounded set of problems. Also this privacy preservation measures can be integrated with the cloud framework and can be extended to Artificial Intelligent and Intelligence Systems like speech recognition, finger print scanner, face detection algorithms etc...

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