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Asian Journal of Computer Science Engineering 2016;1(1):13-16

# **REVIEW ARTICLE**

# **Data Mining**

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Received on: 19/07/2016, Revised on: 25/10/2016, Accepted on: 30/10/2016

#### ABSTRACT

"Computers have permitted us a fountain of wisdom but refused to handle a huge amount of data". To get over this data, the scholars of 1990s introduced the advanced problem solving methodology na med "Data Mining" which has given us abundant of opportunities for exploring and analyzing new key type of data in new ways. In simple, Data Mining is a process of employing one or more computer technique to automatically analyze & extract knowledge from large scale data. For this paper we overviewed, the process model of data mining and its tasks.

Keywords- Data mining, pattern

#### **INTRODUCTION**

The amount of data kept in computer is increasing day by day. And the user are using it for getting more a nd m ore s atisfactory i nformation. A shopkeeper is not happy with just the knowledge of i ts c ustomers, but he t oo w ant t o kno w maximum a bout them. And the S QL l anguages are not enough to fulfill the raising demands for information.

Data mining is working as blessing to solve the problems of pe ople. D ata m ining i s s omething which help us to extract our useful data from huge amount of data.

In a broad s ense, da tam ining i s a f our s tep process to perform a data mining session. We:

- Assemble a collection of data to analyze.
- Present this data to a data mining software program.
- Interpret the result
- Apply t he r esult t o a ne w pr oblem or situation.

At the same instant, Data mining perform many useful tasks.

## **PROCESSING MODEL**

- Assembling the data
- The data warehouse
- Relational database flat files
- Interpreting the result
- Result application

#### Assembling the Data:

To assemble t he da ta, it s hould be acces sed through da ta m ining. D ata c an be a ssembled b y

multiple r ecords in one f ile or s everal f iles. A common misconception is that in order to build an effective model a data mining algorithm must be presented with thousands or millions of instances. In fact, most data mining tools work best with a few hundred or a few thousand pertinent records. Therefore once a problem has been defined, a first step of da ta mini ng p rocess is to extract or assemble a r elevant subset of data for processing. Many times this first step requires a great amount of human time and effort. There are three common ways to access data for mining.

Through Data W arehouse, Relational da tabase and via Spreadsheets.

#### The Data Warehouse:

A common scenario for data assembly shows data originating in one or more operational databases. Operational da tabases a ret ransaction based and frequently d esigned using a relational da tabase model. A nope rational da tabase f ixed on t he relational model will contain several normalized tables. The tables have been normalized to reduce data r edundancy a nd p romote qui ck a ccess t o individual r ecords. F or e xample, a s pecific customer m ight ha ve d ata app earing in several relational t ables w here each table vi ews t he customer from a different perspective.

Data is transferred from operational environment to a da ta w arehouse. A da ta w arehouse is a historical da tabase de signed for de cision s upport rather than transaction processing (Kimball et al., 1996). Thus only data useful for decision support is extracted from the operational environment and

entered into the w arehouse da ta ba se. Data transfer from operational database to warehouse is an ongoing process usually accomplished a daily basis after the close of a regular business day. Before each data items enters into warehouse, the item is time stamped, transformed as necessary, ad checked for errors. The transfer process can be complex, especially when several operational databases are involved. Once entered, the records in the data warehouse become read-only and a re subject to change only under special conditions.

A da ta w arehouse s tores al l da ta r elating t o the same s ubject ( such as customers) i n the s ame table. This distinguishes the data warehouse from an operational database, which stores information so as to optimize transaction processing. Because the data warehouse is subject-oriented rather than transaction-oriented, t he da ta w ill c ontain redundancies. It i s r edundancy s tored n da ta warehouse that i s us ed b y data mining a lgorithm to de velop pa ttern representing di scovered knowledge.

#### **Relational database flat files:**

If a data warehouse does not exist, you can make use of da tabase que ry l anguage s uch as S QL t o write one or more queries to create a table suitable for data mining. From where ever we extract the data we have to transform it int o its r espective data mining tool either its query language or data warehouse. Finally, if a database structure to store the data has not being designed, and the amount of collected data is minimal, the data will likely be stored in a flat file or spreadsheet.

#### Mining of Data:

Prior to giving a data mining tool to a data we have several choices:

• Should 1 earning b e s upervised or un - supervised

• Which instances in the assembled data will be us ed f or bui lding t he m odel a nd which instances will test the model.

• Which attributed will be selected from the list of available attributes.

• Data mini ng tool s r equires the us er t o specify one or m ore l earning pa rameters. W hat parameter setting should be used to build a model to best represent data?

#### **Interpreting the Result:**

Interpreting r esult s ymbolizes us to extract the meaningful data and to find out outcomes of latest discovery of data, If the results are not satisfactory we c an repeat our ex traction process t o get an optimized result.

## **Result Application:**

Our ul timate g oal i s t o a pply what ha s b een discovered to new situations. Suppose through the process of a data mining market analysis we find that t he pr oduct X i s a lmost a lways pur chased with product Y. A classic example of this is the discovery t hat a n unus ually hi gh pe rcentage o f people w ho purchase ba by di apers i n T hursdays also purchase beer. An initial surprise reaction to this f indings m ake s ense w hen we r ealize t hat couples with a young baby at home are not likely to go out on F riday or Saturday night but instead but pr efer t o enjoy t he weekend by relaxing at home. A market analyst can take advantage of this finding b y making beer an obvious di splay items for customers buying diapers.

#### DATA MINING TASKS

It c onsist of t wo m odels: predictive m odel a nd descriptive model as shown in fig 1.1,



Fig 1.1 Data mining tasks

## Classification

Classification m aps da ta i nto pr edefined gr oups or c lasses. It is of ten r eferred t o a s s upervised learning because the classes are determined before examining the data. Two example of classification application are d etermining w hether t o make a bank loan and id entifying c redit r isks. Classification algorithms r equire that the classes be define based on the data attribute values. They often describe t hese cl asses b v l ooking at t he characteristics of data already known to belong to the cl asses. Pattern r ecognition is at ype o f classification where an input pattern is classified into one of several classes based on its similarity to these predefined classes.

#### Regression

Regression is us ed to map a da ta i tem to a real valued pr ediction va riable. i n a ctuality, regression i nvolves the l earning of the f unction that does this mapping. Regression that the target data fit i nto s ome know n t ype o f f unction (e.g., linear logistic, etc) and then determines the b et function of these type that models the given data. Some type of error analysis is us ed to determine which function is "best".

#### **Time Series Analysis**

With *Time Se ries A nalysis*, the value of an attribute is examined as it various overtime. The value us ually are obtained as evenly space time points (daily, weekly, hourly, etc). There are three basic functions perform in time series analysis. In one case, distance measures are used to determine the s imilarity be tween different time s eries. In second case, the structure of the line is examined to de termine (and perhaps classify) its be havior. The t hird a pplication w ould be t o us e t he historical time series plot to predict future values.

## Prediction

Many real-world data mining a pplication can be seen as predicting future data states based on past and current data. *Prediction* can be viewed as a type o f c lassification. (Note: T his is the da ta mining ta sk that bi as different from pr ediction model, although the prediction task is a type of prediction m odel.) T he di fference i st hat prediction is predicting a future state rather than a current state. Here we are referring to a type of application r ather t han t o a t ype o f da ta m ining modeling a pproach, a s d iscussed earlier. Prediction a pplications include f looding, s peech recognition, m achine l earning, and pa ttern recognition. A lthough future v alues m av be predict us ing time series a nalysis or regression techniques.

# Clustering

Clustering is similar except that the groups are not predefine, but r ather de fined b v t he d ata a lone. Clustering is a lternatively r eferred to as unsupervised learning or segmentation. It can be thought of as partitioning or segmenting the data into groups that might or might not be disjointed. The c lustering i s us ually a ccomplished b y determining the s imilarity among the da ta on predefined attribute the mos t similar da ta a re grouped i nto clusters. Since the clusters are not predefined, a dom ain e xpert i s of ten r equired t o interpret the meaning of the created cluster. A special t vpe of clustering i s cal led segmentation. With segmentation a da tabase is partitioned into disjointed groupings of s imilar tuples cal led segments. S egmentation is of ten viewed as being identical to clustering. In other circles segmentation is viewed as a specific type of clustering applied to a database itself.

## Summarization

*Summarization* maps da ta i nto s ubsets w ith associated simple discriptions. Summarization is also called cha racterization or generalization. It extracts or de rives r epresentative i nformation about the database. This may be accomplished by © 2015, AJCSE. All Rights Reserved.

actually r etrieving p ortions of t he da ta. Alternatively, summary type information (such as the m eaning of s ome n umeric at tribute) c an be derived f rom the da ta. The s ummarization succinctly ch aracterizes t he cont ents of t he database.

## **Association Rules**

*Link analysis*, alternatively referred to as *affinity* analysis or association, refers to the data mining task of uncovering relationships among data. The best e xample of this type of a pplication is to determine association rules. An association rule is a mode 1 that ide ntifies s pecific t ypes of d ata association. These as sociations are often used in retail s ales c ommunity to identify ite ms that a re frequently pur chased t ogether. A ssociation a re also us ed i n m any ot her a pplications s uch a s predicting the f ailure of te lecommunication switches. U sers of a ssociation r ules m ust b e cautioned t hat t hese a re not c asual r elationships. They do not represent any relationship inherent in the act ual da ta ( as i st rue w ith functional dependencies) or in the real world. There probably is no relationship between bread and pretzels that causes them to be purchased together. And there is no guarantee that this association will apply in the future. However, association rules can be used to assist r etail s tore m anagement i n effective advertising, marketing and inventory control.

## **Sequence Discovery**

Sequential analysis or sequence discovery is used to determine s equential patterns in data. These patterns are based on a time sequence of actions. These patterns are similar to associations in the data (or events) are found to be related, but the relationship is based on time. Unlike the market analysis, which requires the items to be purchased at the same time, in sequence discovery the items purchased over time in some order. A similar type of di scovery c an b e s een in the s equence within which da ta a re pur chased. F or e xample, m ost people pur chase C D pl ayers m ay be f ound t o purchase C Ds within one week. As we will s ee, temporal a ssociation rules r eally f all int o this category.

## CONCLUSION

Today's pe ople is drowning in da ta but starving for know ledge. Therefore D ata m ining i nvolves extracting meaningful information, rules, patterns from hu ge d eposited da ta. In t oday's e ra m any useful t echniques ar e a vailable t o extract da ta. Data mining is a method to search hidden data and to acknowledge the entities.

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