(I)Date mining tasks clarified into two catagories

Clastering, Asscriptive orining fasks

(1) Descriptive - Charocharize the general proposettes

of the data in the database. (2) pradictive - predictive mining tasks paretorm inference on the current data Time comes Analysis, predictions one are to make predictions. Hamilication Regnession. Date mining functionalities (I) Characterization and Discrimination. Charcoctenization -> scommarizing the darty of the day enden study (torget com) Discrimination > comparison of the tanget class with one or a set of comparative class. Mining frequeent postorers, 4550 ciahan and concertions. Frequent patterns > patterns that occur frequently Mining frequent patterns help to discover the associations and counsenting nathin dute. Association analyphy buys (x, "computer") > buy 8(x, "spo") [surport=1x confidence of 50% -> if a customan buys a computer ,- liere is a 50% chance that he/she will buy S/10 as wall. 14. support of 14. of all the françactions enden anonyris mowed that pead S/W were porchared togetier. buy > is the attribute. age (x, "20...29") Nincome (x, "20K...-29K")=> buy 8 (x, "ed player") [Support 2%, confidence. Got] out of all customers and an study 2% one 20 to 29 yrs. of age with an income of 2014-2914 and here punchased a co playon.

There is 60% probability that a customer in this age and income group will possehole a Coplayor.

> Association between more than one attributes
entry age, income, and buys.

clamification and prediction.

that describes and distinguishes data crosses for the purpose of being able to we the model to predict the class of objects whose cross label is wirning.

The decived model is based on the enalyting of a set of training date.

consision trees or neural network

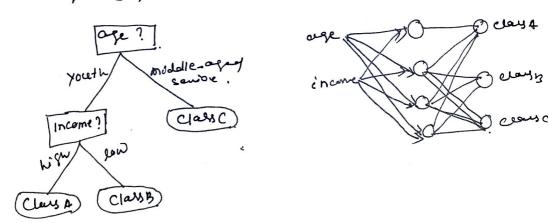
(a) age (X, "youth") AHD income (X, "Low") > coals(X, "A")

age (X, "youth") AHD income(X, "low") > coals(X, "B")

age (X, "youth") AHD income(X, "low") > coals(X, "c")

age (X, "middle_aged) > coals(X, "c")

age (X, "serior")



prediction > gt is used to predict mining on marailable numerical data value

statistical methodology regression analysis
is most oftening used for numeric prediction.

Out fion analyting

A database may conterin date Objects that do not comply we'the the seneral Schenice or model of the data. The analysis of outline moning.

out lions may be detected using. > statistical tests that wind probability moder.

-> using distance measures

deviation based methods = cidentity

outliens by exemining differences

in the main characterishes of objects

in a presup.

Er 9+ may crawers fraudulant wrote of credit cands by detecting purchases of extremely early amounts force given account no- c'n companison to requen Charges incurred by the same occount.

Evolution analytis.

Dat evolution analysis describes and modely trends for objects where behavior changes over time

Ex stock market forecasting.

(2) According to the kinds of knowledge mored.

> they use thee data ming functionalities.

(3) According to the kinds of techniques

whilized.

Ere > structure learning, sattistics, partern

recognition or HH.

(4) According to thre applications adapted. Ext finance, telecommendin, DNA, Story montest etc.

Data mining lask primitives

form of a data mining quary, which is input to the data mining system.

A deta mining query is defined in teams of data mining task primitives.

The data mining primitives specify the followings: The data mining tack course specified by 5 primitives:

- (1) The set of fusk-relevant data to be mined.

 This specifies the postrons of the database

 or the set of data in which the user is

 in terested. Ex:- relevant attributes on dimensions.
- (2) The kind of knowledge to be mined.

 This specifies the data mining functions to be performed, such as characterization, discrimination, association, crawification etc.

Jata mining primitive slatine a seater mining job, which combe specified in the found of a slate mining queny.

- discovery process.
 - The know redge about the demain to be moned its useful for quiding the knowledge discovery process and evaluating the pasterns.
- (4) The interestingness measures and thresholds

Decreenty (voledity)

Decreenty (voledity)

Butlety (week They may be used to quide the monong

Butlety (week)

Provery process or after discovery, to evaluate the discovered patterens.

- (5) The expected representation for visualizing. the discovered patterns.
 - > refers to the form in which discovered patterns are to the presented, for example on the form of rules, falsoes, chants etc.

Integration of Data mining system with a dætabase, or dæter wærehoure system.

- DM system will not use (1) No coupling > any function of a DB/DW system. of uses flat filles as data sources.
- (2) Luose coupling > DM system we'll use some forcidities of a DB OR BW sustem.
- losse coupling is better han no wurling because it can peter any postion of data Stoned in databases or data wanchouses by using query prescentry, indexing and other system Lociedties.

-> 94 is difficult for loose coupling to enchieve hugh scalability and good pereformance with sange data sets.

semitight coupling.

97 mans berides linking a DM system to a DB/DW system, efficiens implomentation of a frew essential data moning primitives can be previded in the DB/DW system.

- > These presonitives include soleting, indexing, aggregation, histogram analysis extente
- > Because these interemediate mining results are either precumputed or con be computed efficiently, this design ue'll enhance the pereforemencie of a DM system.

Tight coupling 9+ many a DIU System i's i'ntegrated into the DB/DW system.

majore d'escres d'n Data mining.

- (1) Mining different kinds of knowledge indulesak >> Recourse different usens one interested of I roccourse differents of knowledge, data mining different kinds of knowledge, data mining should cover a vide rounge of data analysing and knowledge discovery tasks.
 - Is The datas maining tasks may use the same database in different ways and arguing the development of numerous dute moning te chiques.