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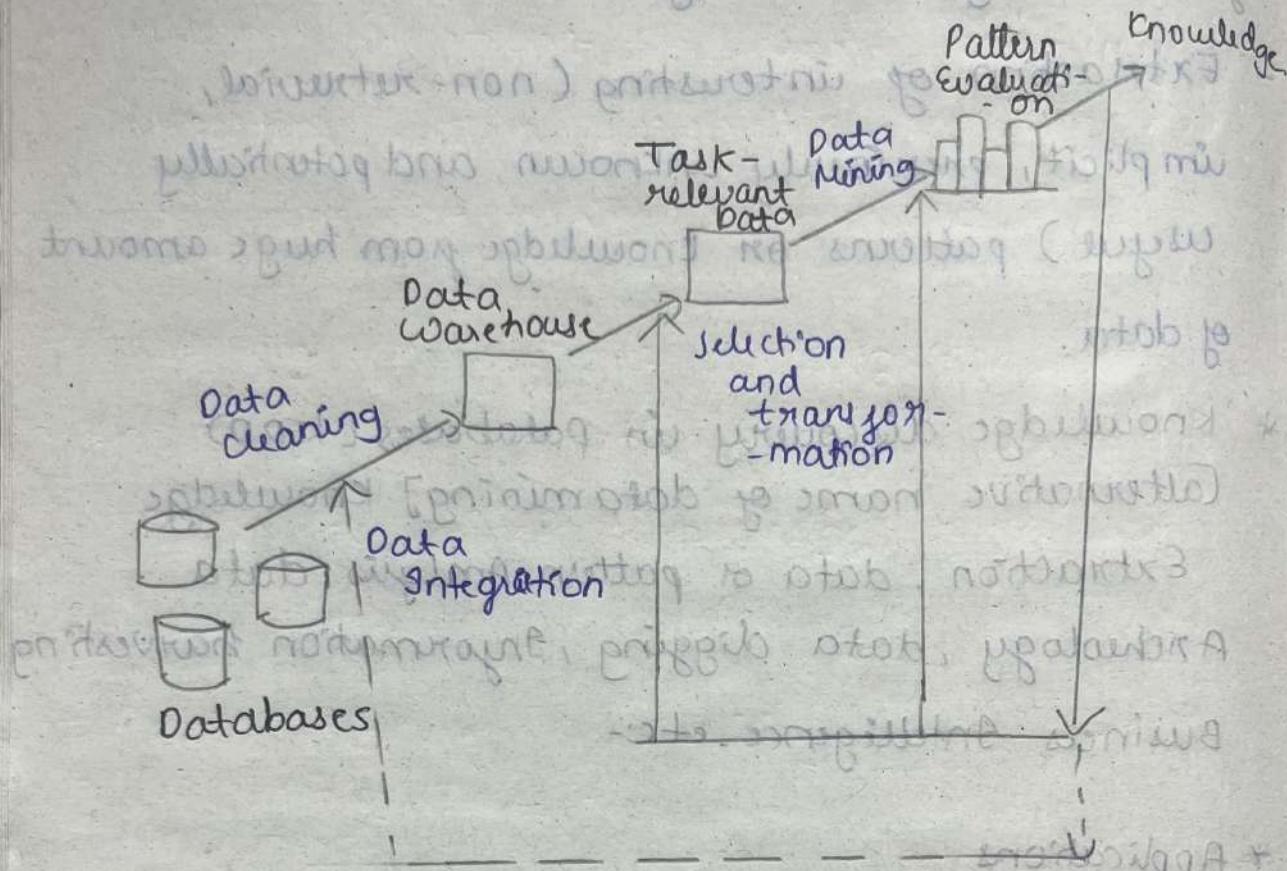
UNIT-I

- * What is Data Mining :- Extracting information from huge amount of data (a)
- Extraction of interesting (non-retrivial, implicit, previously unknown and potentially useful) patterns or knowledge from huge amount of data.
- * Knowledge discovery in Databases (KDD)
[alternative name of data mining] knowledge extraction, data or pattern Analysis, data Archaeology, data digging, information harvesting, Business Intelligence, etc:-

* Applications

- Database Technologies
- pattern recognition
- Data visualization
- statistics
- Algorithms
- etc:-

* Data Mining - core of knowledge discovery process



* Data Mining on what kinds of data.

- Data warehouses
- Relational databases
- Traditional databases
- Advanced database
- Information repositories
- Object Oriented and object Relational databases
- Spatial Database
- Time series data
- Temporal data
- Text database
- Multi media database.

- heterogeneous and legacy databases
- World wide web, etc.
- Data Mining Applications:-
 - * Customer segmentation :- Business use data mining techniques to understand customers.
 - * Market basket Analysis :- In Retail out of 10 customers, if 8 are purchasing two items or both items then both will be kept near.
 - * Risk management :- In LIC's, Banks Loans.
 - * Fraud detection :- In Banks to take loan.
 - * Demand Prediction.

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- KDD steps (process)

- Data cleaning:- To remove noise and inconsistent data (may take 60% of effort).
- Data integration:- when multiple data sources may be combined.
- Data selection:- where data relevant to the analysis ^{task} are retrieved from the database.
- Data transformation:- where data are transformed or consolidated into forms appropriate for mining by performing summary or aggregation.

- Data Mining :- search for patterns of interest
An essential process where intelligent methods are applied in order to extract data patterns.
- Pattern evaluation:- To identify the truly interesting pattern representing knowledge based on some interestingness measures.
- Knowledge presentation:- visualization and knowledge representation techniques are used to present the mined knowledge to the user.

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- Architecture of Data Mining
- Data Mining Tasks :-
- Predictive :-
- It is a supervised learning from data
- Predictive values of data by making use of known results from a different set of sample data.

Predictive

→ Classification

→ Regression

→ Prediction

→ Time series

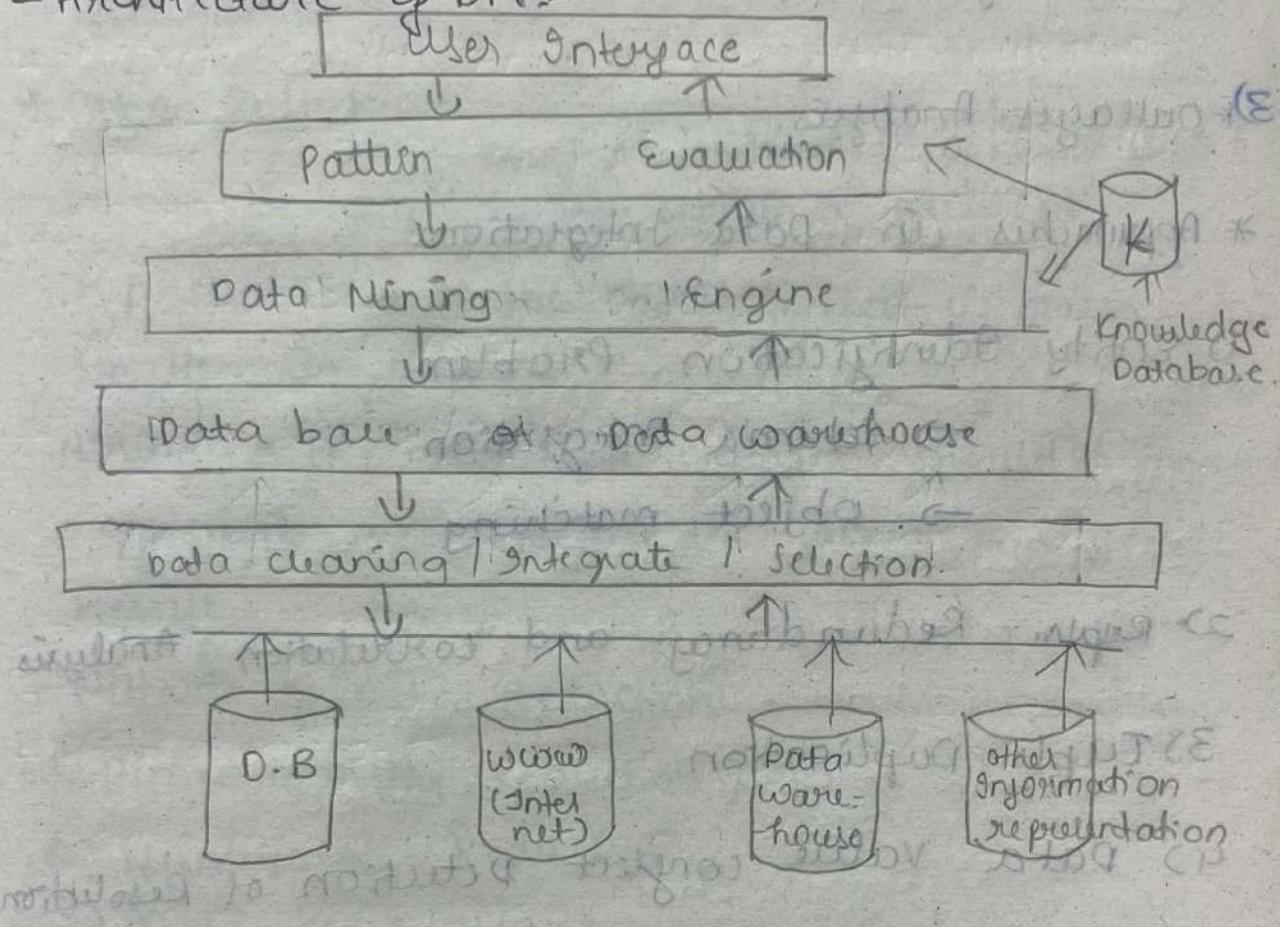
- Descriptive :-

- Enables us to determine patterns and relationship in a sample data

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Descriptive
 → Association rules
 → Clustering
 → Sequence discovery
 → Summarization.

- Architecture of DMI:

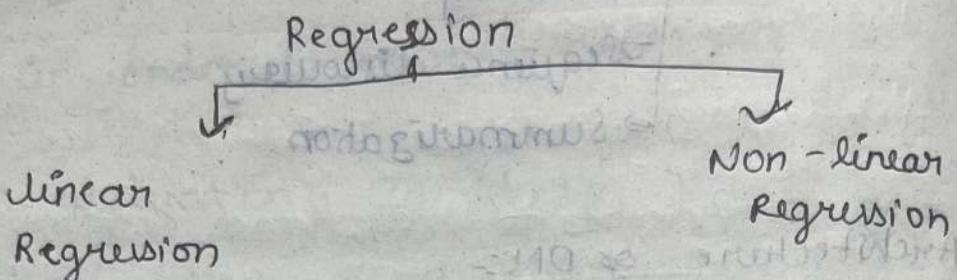


* Data cleaning :- whatever the noisy data that cleans by using filling the missing values, smoothening noisy data, resolving the inconsistency and removing the outliers.

* Binning :- Partitioning into equal sets, smoothening the bins by replacing by mean or average of null values. Smoothing by bin boundaries.

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2) Regression :-



3) Outliers Analysis

* Approaches in Data Integration.

1) Entity Identification Problem

→ Schema integration

→ object matching

2) Repeating Redundancy and correlation Analysis

3) Tuple Duplication

4) Data value conflict Detection or Resolution

* Data Integration :-

- Merging of data collected from multiple sources,

careful integration can help reduce redundancies, and inconsistencies in the resulting dataset.

- Tuple duplication.

- The use of denormalized tables (often done to improve performance by avoiding joins)

- Denormalized table is another source of data redundancy.
- Inconsistency often arises arise between various duplicate due to the inaccurate data entry or updating sometimes, but not all data occurrences

* Data Reduction :-

- 1) Dimensionality Reduction
 - 2) Numericity Reduction
 - 3) Data compression
- It can be applied to obtain a reduced representation of the dataset, that is much smaller in volume.
 - Mining on the reduced dataset should be more efficient and produce the same analytical result.
 - Methods of Data Reduction:-
 - 1) Dimensionality Reduction (DR)
 - 2) Numericity Reduction
 - 3) Data compression

* DR :- It eliminates the redundant attributes which are weakly important across the data.

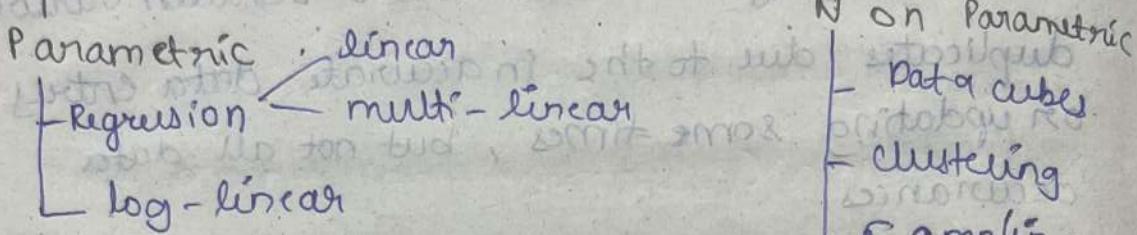
1) Step wise forward selection

2) Step wise backward elimination

3) Decision tree induction.

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* Numerosity Reduction (NCR)



* Numerosity reduction techniques

- 1) Parametric
- 2) Non Parametric.

* Parametric methods:- In this data is represented using some model.

- In this, we need to estimate the data so that only parameters of data are required to be stored instead of actual data.
- The two different methods are:-

1) Regression

2) Log-linear method.

- Used for creating such models
- Regression:- Regression can be a simple linear regression or multi-linear regression, when there is only single independent attribute, such model is called simple regression.
- In this the data are modelled to fit a straight line.

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* Multi-linear Regression:- there are multiple independent attributes used such regression models are called Multi-linear Regression.

- log-linear model:- In this model used to estimate the probability of each data point in a multi-dimensional space, for a set of discrete attributes based on a smaller set of dimensional combinations

* Non Parametric methods:- This methods are used for storing reduced representations of data include histograms, clustering, sampling and data cube aggregation.

- histogram:- This is the data representation in terms of frequency.

- clustering:- This technique partitions the whole data into different groups of clusters.

- Sampling:- This can be used for data reduction because it allows a larger dataset to be reduced? represented by a much smaller random data sample. (Subset of original data).

- Data cube aggregation:- It involves moving the data from detail level to a ^{fewer} _{pure} number of dimensions, the resultant data set is smaller in volume without loss of info, necessary further analysis starts task.

* Data compression:- In this modification, encoding or converting, the data structure of data in a way that consumes less space.

- we can divide it into two types based on their compression techniques
1) lossless compression
2) lossy compression.

1) lossless compression:- Encoding techniques allow a simple and minimal data size reduction.
- It uses algorithms to restore the precise original data from the compressed data.

2) lossy compression:- The compressed data may be differ to the original data, but are useful enough to retrieve info from them.
- They are
1) Discrete wavelet transforms
2) Principle component analysis.

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- Discrete wavelet transformation:- This is a signal processing technique that transforms linear signals.
- The wavelet transforms the data can be concatenated and this is helpful in data reduction.

- If we store a small fraction of a strongest wavelet coefficient then the compressed approximation of the original data can be obtained.
- Example:- An image of size 100 MB compressed to 100 KB. Here picture quality decreases.
- Principle component analysis(PCA):- In this analysis we are extracting the important variables from a large number of variables available in a data set, it extracts a set of low dimensional features from a high dimensional dataset with a goal of capturing as much information as possible in data.

* Steps involved in

- 1) Standardize the dataset
- 2) Compute the covariance matrix for features in data sets.
- 3) Compute the Eigen values and Eigen vectors.
- 4) For the covariance matrix.
- 5) Sort the Eigen values and their corresponding Eigen vectors. Choose K Eigen values to form an Eigen vector matrix.
- 6) Transform the original matrix.