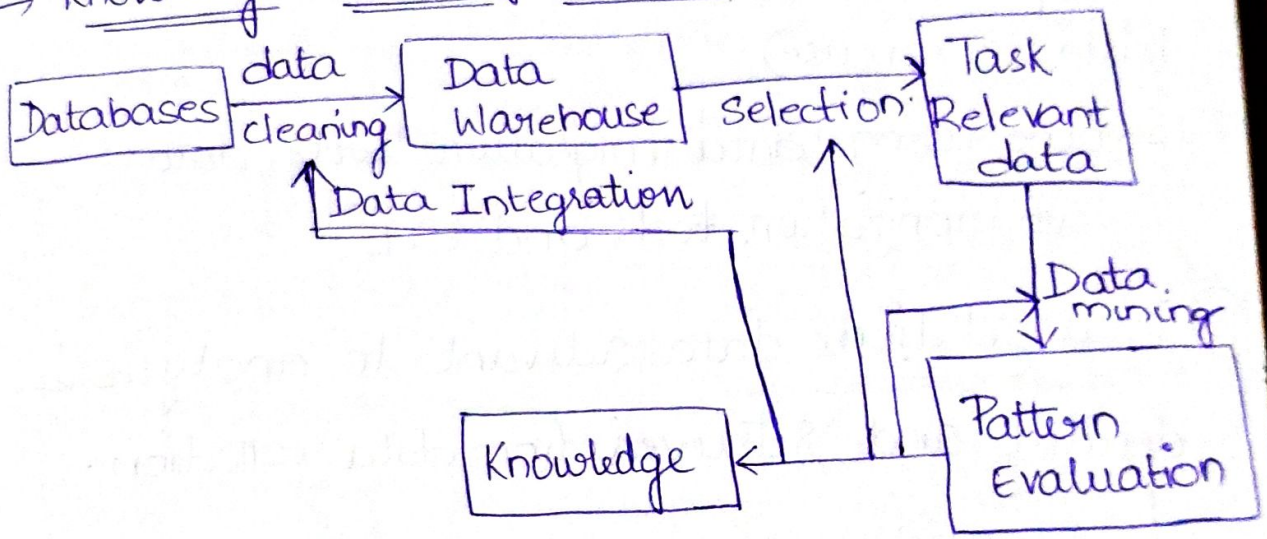
27/09/2024

UNIT - 1

→ Data mining is the extraction of interesting patterns or knowledge from huge amount of data.

→ Data mining is also called knowledge discovery in databases, knowledge extraction, business intelligence etc.

→ Knowledge Discovery Process: (KDD)



→ Ex: Web Mining Steps:

↳ Data cleaning, data integration, warehousing the data, data cube construction, data selection, data mining, presentation of results, knowledge.

* Steps included in KDD process:

① Data cleaning: removal of noisy and irrelevant data from collection.
- ↳ missing values
- ↳ noisy data
- ↳ data discrepancy detection & data transformation tools.

② Data integration: heterogeneous data from multiple sources combined in a common source (data warehouse).
- ↳ done using data migration tools, data synchronization tools and ETL

③ Data selection: data relevant to analysis is decided and retrieved from data collection.
- ↳ done using neural network, decision trees, Naive Bayes clustering & regression methods.

④ Data transformation: transforming data into appropriate form.
- ↳ 2 steps:
  - → Assigning elements from source to destination : Data Mapping
  - ↳ Code generation: transformation program.

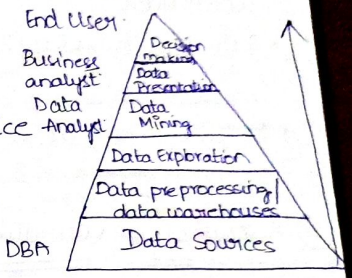
⑤ Data Mining: extract patterns potentially
- ↳ decided using classification con characterization

⑥ Pattern Evaluation: identifying strictly ↑sing patterns representing knowledge based on given measures.

⑦ Knowledge Representation: score of each pattern and uses summarization & visualization.

* Advantages of KDD:

① Used efficiency
② better customer service
③ fraud detection
④ Predictive modelling.



* Multi-dimensional view of data mining:

→ Data to be mined: database data.

→ Knowledge to be mined: characterization, discrimination, association, classification etc.

→ Techniques used: Data warehouse (OLAP), ML, statistics, pattern recognition etc.

→ Applications adapted: telecommunication, banking, fraud analysis etc.

* Kinds of data: data streams, sensors data, time-series data, temporal data, sequence data etc. Multimedia, Text, Object-relational databases etc.

* Functions of data mining:

① Generalization:
* Information integration and data warehouse construction.
* Data cube technology: Online Analytical processing (OLAP).
* Multi-dimensional concept description: Characterization & discrimination.

② Association & Correlation analysis:
* Frequent patterns.
* association, correlation v/s causality.

③ classification:
* Classification & label prediction.
  - → construct models.
  - → describe & distinguish classes (or) concepts
  - → predict some unknown class labels.

* Typical methods:
  - → decision trees, support vector machines
  - → Neural networks, pattern based classification

* Applications:
  - → Fraud detection, classifying stars

④ Cluster Analysis:
- → unsupervised learning
- → group of data form new categories.
- → maximizing intra-class similarity & minimizing inter-class similarity

⑤ Outlier Analysis:
- → Outlier: a data object that does not comply with general behaviour of the data.
- → methods: product of clustering, regression analysis.

* Evaluation parameters:
- → descriptive v/s predictive
- → coverage → Accuracy → Timeliness
- → Typicality v/s novelty.

* Confluence of Multiple Disciplines:
- → ML, Pattern recognition, Statistics, applications, algorithm, database technology, HPC, visualization.
- → used due to.
  1. Tremendous amount of data
  2. High dimensionality of data.
  3. High complexity of data.
  4. New & sophisticated apps.

* Applications of DM:

→ Web page analysis
→ Collaborative analysis
→ Basket data analysis to targetted marketing
→ Biological & medical data analysis

* Issues in data mining:

① Mining Methodology - handling noise, uncertainity
② User interaction - visualization & user interaction
③ Efficiency & scalability - parallel, distributed, stream
④ Diversity of data types - complex data types
⑤ Data mining and society - impacts of DM, privacy

* Types of data sets:

→ Record: Relational records, data matrix. etc.
→ Graph and network: WWW, molecular structures.
→ Ordered: video data, genetic sequence data.
→ Spatial, image and multimedia: image data, video data.

* Important characteristics of structured data:

① Dimensionality
② Sparsity
③ Resolution
④ Distribution.

→ Data object represents an entity.
→ Data objects are described by attributes.
→ Ex: sales database, medical database
→ Attribute: a data field, representing a characteristic or feature of a data object.

Types:

① Nominal: categories, states, names of things
Ex: marital status, occupation, zip codes.

② Binary: 2 states = 0/1
↳ Symmetric: both becomes equally important.
Ex: gender.
↳ Asymmetric: outcomes not equally important.
Ex: medical test (+ve v/s -ve)

③ Ordinal: values have a meaningful order but magnitude b/w successive values is not known.

→ Numeric Attribute Types:
(Integer (or) real-valued)

* Interval: measured on a scale of equal-sized units.
Ex: Temp in °C or °F, dates

* Ratio: inherent zero-point.
Ex: Temp in K, length, counts.

→ Discrete v/s continuous attributes:

① Discrete: Has only a finite (or) countably infinite

set of values (Integer variables)
Ex: zip codes, profession.

② Continuous attributes: Has real numbers as attribute values. (float-pointing
Ex: temperature, height (or) weight.

* Basic Statistical descriptions of data:

→ central Tendency, variation and spread.
→ median, max, min, quantiles, outliers, variance etc.
→ Boxplot (or) quantile analysis on sorted intervals as well as transformed cube.

** Mean: $\bar{x} = \frac{1}{n}\sum_{i=1}^{n} x_i \qquad \mu = \frac{\sum x}{N}$

Weighted arithmetic mean: $\bar{x} = \frac{\sum_{i=1}^{n} w_i x_i}{\sum_{i=1}^{n} w_i}$

** Median:

$$= L_1 + \left(\frac{n/2 - (\sum \text{freq})l}{\text{freq}_{\text{median}}}\right) \times \text{width}$$

** mode = 3 median − 2 mean.

** variance $\Rightarrow s^2 = \frac{1}{n-1}\sum_{i=1}^{n}(x_i - \bar{x})^2$

$$\sigma^2 = \frac{1}{N}\sum_{i=1}^{n}(x_i - \mu)^2 = \frac{1}{N}\sum_{i=1}^{n} x_i^2 - \mu^2$$

* Properties of normal distribution

→ from $\mu - \sigma$ to $\mu + \sigma$: about 68%.
→ from $\mu - 2\sigma$ to $\mu + 2\sigma$: contains about 95%.
→ from $\mu - 3\sigma$ to $\mu + 3\sigma$: contains about 99.7%.

* Graphic displays of Basic Statistical descriptions:

→ Boxplot: graphic display of 5-number summary.
→ Histogram: x-axis: values, y-axis: representation frequencies.
→ Quantile plot: each x_i is paired with f_i
→ Quantile-quantile (q-q) plot: graphs the quantiles of one univariant distribution against the corresponding of another
→ Scatter plot: plotted as points in the plane.

* Data visualization:

→ Importance:
** Gain insight
** Provide qualitative insight
** Search for patterns
** Help find suitable regions & parameters.
** provide visual proof.

→ Categorization of visualization methods:

1. Pixel-oriented visualization techniques.
2. Geometric projection "
3. Icon-based "
4. Hierarchical "
5. Visualizing complex data & relations.

(1) → Income, Credit Limit, transaction volume, age.

(2) methods: direct visualization, scatterplot & scatterplot matrices, prosection views, hyperslice, parallel coordinates.

(3) methods: Chernoff faces, Stick figures.
general: shape coding, color icons, tile bars.

(4) methods: dimensional stacking, tree map, cone trees, info cube.

* Similarity and dissimilarity:

→ Similarity: Numerical measure of how alike 2 data objects are

range = [0,1]

→ Dissimilarity: Numerical measure of how different 2 data objects are.

• min. dissimilarity = 0 ; upper limit varies.

→ proximity refers to a similarity (or) dissimilarity

* Data Matrix:
  - n data points with p dimensions.
  - 2 modes.

* Dissimilarity Matrix:
  - n data points, but registers only the distance
  - triangular matrix
  - single mode.

* Proximity Measure for nominal attributes:

→ can take 2 (or) more states.

Method-1: Simple Matching

$$d(i,j) = \frac{p-m}{p}$$

Method-2: large no. of binary attributes

→ A contingency table for binary data:

	1	0	sum
1	q	r	q+r
0	s	t	s+t
Sum	q+s	r+t	p

* distance measure for

→ symmetric: $d(i,j) = \frac{r+s}{q+r+s+t}$.

→ asymmetric: $d(i,j) = \frac{r+s}{q+r+s}$.

→ Jaccard coefficient = $\frac{q}{q+r+s}$

* Dissimilarity blw binary variables:

Ex:

Name	Gender	Fever	Cough	Test-1	Test-2	Test-3	Test-4	
Jack	M	Y	N	P	N	N	N	P=1
Mary	F	Y	N	P	N	P	N	Y=1
Jim	M	Y	P	N	N	N	N	N=0

$d(\text{Jack, Mary}) = \frac{r+s}{q+r+s} = \frac{0+1}{2+0+1} = \frac{1}{3}$.

$d(\text{Jack, Jim}) = \frac{r+s}{q+r+s} = \frac{1+1}{1+1+1} = \frac{2}{3}$.

$d(\text{Jim, Mary}) = \frac{r+s}{q+r+s} = \frac{1+2}{1+1+2} = \frac{3}{4}$

* Standardizing Numeric Data:

$z = \frac{x - \mu}{\sigma}$

x = raw score to be standardized.

μ = mean of population

σ = standard deviation.

→ minkowski distance; (Slide -58)

$$d(i,j) = \sqrt[h]{(x_{i1}-x_{j1})^h + (x_{i2}-x_{j2})^h + \dots + (x_{ip}-x_{jp})^h}$$

properties:

→ $d(i,j) > 0$ & $d(i,i) = 0 \Rightarrow$ +ve definiteness

→ $d(i,j) = d(j,i) \Rightarrow$ Symmetry

→ $d(i,j) \leq d(i,k) + d(k,j) \Rightarrow$ Triangle Inequality

→ Manhattan distance:

$$d(i,j) = |x_{i1}-x_{j1}| + |x_{i2}-x_{j2}| + \dots + |x_{ip}-x_{jp}|$$

→ Euclidean distance:

$$d(i,j) = \sqrt{(x_{i1}-x_{j1})^2 + (x_{i2}-x_{j2})^2 + \dots + (x_{ip}-x_{jp})^2}$$

* Cosine Similarity:

$$\cos(d_1, d_2) = \frac{(d_1 \cdot d_2)}{|d_1| \cdot |d_2|}$$

Ex: $d_1 = (5,0,3,0,2,0,0,2,0,0)$

$d_2 = (3,0,2,0,1,1,0,1,0,1)$

$\cos(d_1, d_2) = 0.94$

$d_1 \cdot d_2 = 5*3 + 0\times0 + 3\times2 + \dots + 0\times1$

$= 25$.

$|d_1| = \sqrt{5\times5 + 0^2 + 3^2 + \dots + 0^2} = 6.48$

$|d_2| = \sqrt{3\times3 + 0^2 + 2^2 + \dots + 1^2} = 4.12$

→ Z-Score & Min-Max Normalization:

data = 1000, 2000, 3000, 5000, 9000

Min = 0, Max = 1.

min = 1000 max = 9000.

$$v = \frac{x - min}{max - min} \Rightarrow$$

0
0.125
0.25
0.5
1

$$Z = \frac{x - \mu}{\sigma}$$

$$\mu = \frac{20000}{5} = 4000.$$

$$\sigma = \sqrt{\frac{\Sigma(x_i - \mu)^2}{n-1}}$$

$$= \sqrt{\frac{(1000-4000)^2 + (2000-4000)^2 + \cdots (9000-40}{4}}$$

$$= 2489.97$$

-1.204
-0.803
-0.4016
0.4016
2.008