

IT402: SOFT COMPUTING

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1. Hard computing

It requires programs to be hard written.

It uses two valued logic. It has features of exactness and categoricity.

It is deterministic in nature.

It works on clean and exact data.

It produces exact results.

It produces settled code and results.

Soft computing

It can be evolved into its own programs.

It uses fuzzy logic. It has the features of approximation and dispositionality.

It is stochastic in nature.

It works on ambiguous, dirty and noisy data.

It produces approximate results.

It incorporates randomness.

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Sequential computation	Can perform parallel computation.
For Exactly stated analytical model.	Tolerant to imprecise, approximation-based and uncertain data.
eg:- Merge sort, Search algorithms	eg:- Fuzzy logic, neural network.

Components of Soft Computing

1. Probabilistic Logic:- It handles the uncertainty to use formal argument structure.
2. Machine Learning:- used to improve performance on a task by remodeling.
3. Evolutionary computation:- It is a subfield of artificial intelligence, used for global optimisation.
4. Fuzzy Logic:- handles partial truth that can range between 0 and 1.
5. Neural Networks:- a collection of neurons forming layers to solve a problem.
6. Support Vector machine:- used for classification and regression.

7. Hybrid techniques can be used to achieve better results in soft computing

8. Genetic Algorithm

Goals of soft computing

Main goal:- develop intelligent machines to solve real world problems which are difficult to model mathematically.

Resemblance with human decision making

Exploit the tolerance for uncertainty, imprecision, approximation and partial truth.

Rapid dissemination of important results through soft computing

Image processing and data compression

Data mining

Power and control systems

Handwriting recognition

Clustering based on similarity

Face recognition and speech recognition

2.

$$k = 5$$

age ≤ 30 , income = medium, student = yes,
credit-rating = fair.

ID	Similarity (A, B)
1	$(1/4 + 0 + 0 + 1/4) = 1/2 = 0.5$
2	$1/4 + 0 + 0 + 0 = 1/4 = 0.25$
3	$0 + 0 + 0 + 1/4 = 1/4 = 0.25$
4	$0 + 2/4 + 0 + 1/4 = 3/4 = 0.75$
5	$0 + 0 + 1/4 + 1/4 = 2/4 = 0.5$
6	$0 + 0 + 1/4 + 0 = 1/4 = 0.25$
7	$0 + 0 + 1/4 + 0 = 1/4 = 0.25$
8	$1/4 + 2/4 + 0 + 1/4 = 4/4 = 1$
9	$1/4 + 0 + 1/4 + 1/4 = 3/4 = 0.75$
10	$0 + 2/4 + 1/4 + 1/4 = 1$
11	$1/4 + 2/4 + 1/4 + 0 = 1$
12	$0 + 2/4 + 0 + 0 = 0.5$
13	$0 + 0 + 1/4 + 1/4 = 0.5$
14	$0 + 2/4 + 0 + 0 = 0.5$

Since $k = 5$

IDs considered: 4, 8, 9, 10, 11

majority of IDs have class label "Yes"

Hence, the new example with attributes
age ≤ 30 , income = medium, student = yes,
credit-rating = fair is "Yes"

3.

Agglomerative

More informative than unstructured set of clusters.

Bottom-up approach

Less efficient as time complexity $O(n^2)$ after optimisation.

Doesn't take into account global distribution of data.

Starts with every data point as ~~se~~ a cluster

Divisive

More complex

Needs a flat clustering method as subroutines

Top-down approach

More efficient $O(n)$ mostly

More accurate. It takes global distribution of data

Starts with all data points as one cluster

Agglomerative method

	1, 1	1, 4	1, 6	5, 1
(1, 1)	0	3	5	4
(1, 4)	3	0	2	3+4=7
(1, 6)	5	2	0	9
(5, 1)	4	7	9	0

~~1, 1~~

1, 4 ——— (1, 4), (1, 6)

1, 6 ———

1, 1 ———

5, 1 ———

((1, 1), (5, 1))

P_1, P_2, P_3, P_4

4.

$$\text{Info}(D) = -\frac{5}{14} \log_2\left(\frac{5}{14}\right) - \frac{9}{14} \log_2\left(\frac{9}{14}\right)$$

$$= -0.159 - 0.123$$

$$= 0.2823$$

$$\text{Info}_{\text{outlook}}(D) = \frac{5}{14} \log_2\left(\frac{5}{14}\right) + \frac{9}{14} \log_2\left(\frac{9}{14}\right)$$

1, 2

2, 1

3, 1

4, 1

A

B

C

D

(1, 1)

F = A + B

C

D

E

(A, 1)

F

G

H

I

(2, 1)