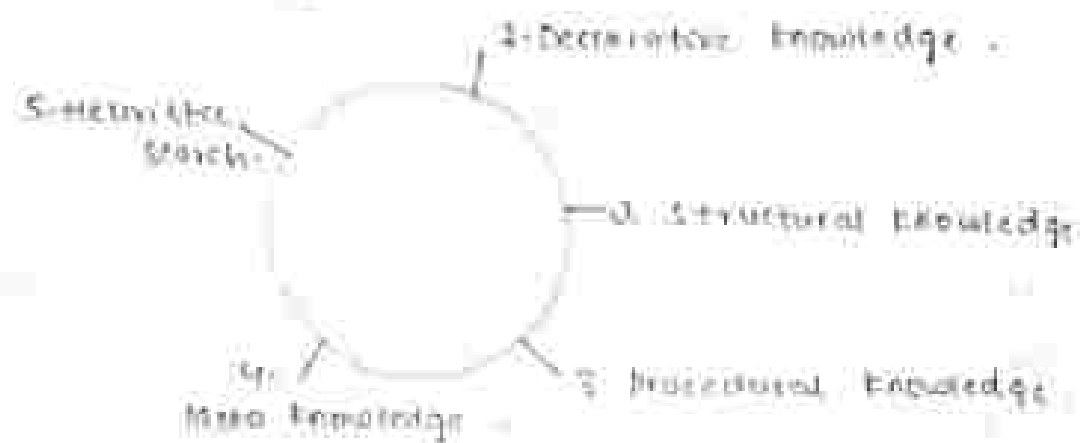


Unit 9 - Knowledge Representation

Knowledge Representation represents the information from the real world for a computer to understand and then utilize the knowledge to solve complex problems, like communicating with human beings in natural language.

Knowledge Representation means presenting the knowledge that AI system should acquire in different forms:

Types of Knowledge Representation:



1. Declarative Knowledge

2. Structural Knowledge

3. Procedural Knowledge

4. Meta Knowledge

5. Heuristic Search

1. Declarative Knowledge:

This includes concepts (related to mathematics), objects and facts expressed in declarative sentence:

Ex: All go to Curriculum and academics and respect
CSE-B is Best.

Here Concepts = Curriculum, academics

Object = CSE-B.

1. Structural Knowledge

Basic problem solving knowledge that describes the relationship between concept and object.

Eg. legs, hands, eyes \rightarrow human.

2. Procedural Knowledge

This includes rules, strategies and procedures etc. In water jug problem, we have strategies.

1. Empty

2. Fill and

3. Transfer

and the condition is ~~the~~ no jug is overflowed. And by using these strategies and rules an agent performs process.

3. Meta Knowledge

Meta knowledge contains knowledge other than structural and procedural knowledge.

Eg. other than teaching teacher doing attendance, admission etc. comes under meta knowledge of a teacher.

4. Heuristic Knowledge

This is based on expert's knowledge. It represents initial and final states.

Semantic Network

It is a graphical representation for representing knowledge in patterns of interconnected nodes.

There are 2 types of semantic networks.

1. Simple semantic network.

2. Propositional semantic network.

1. Simple Semantic network

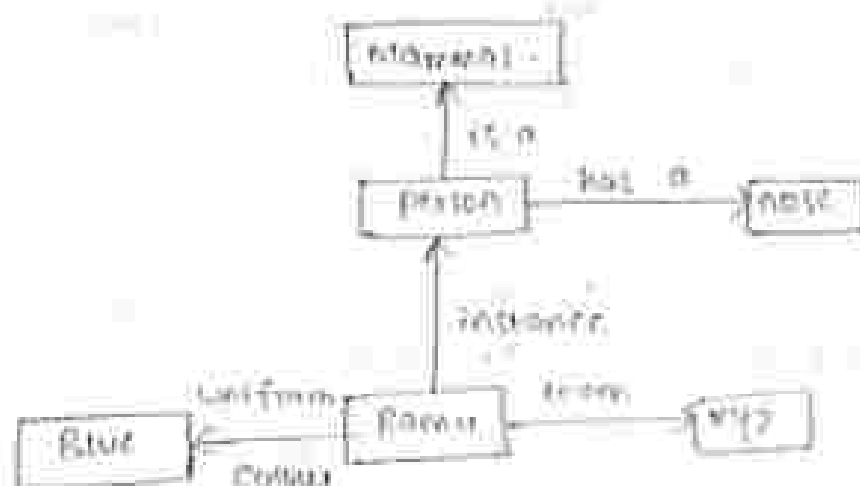
It consists of the following things:

has a

is a

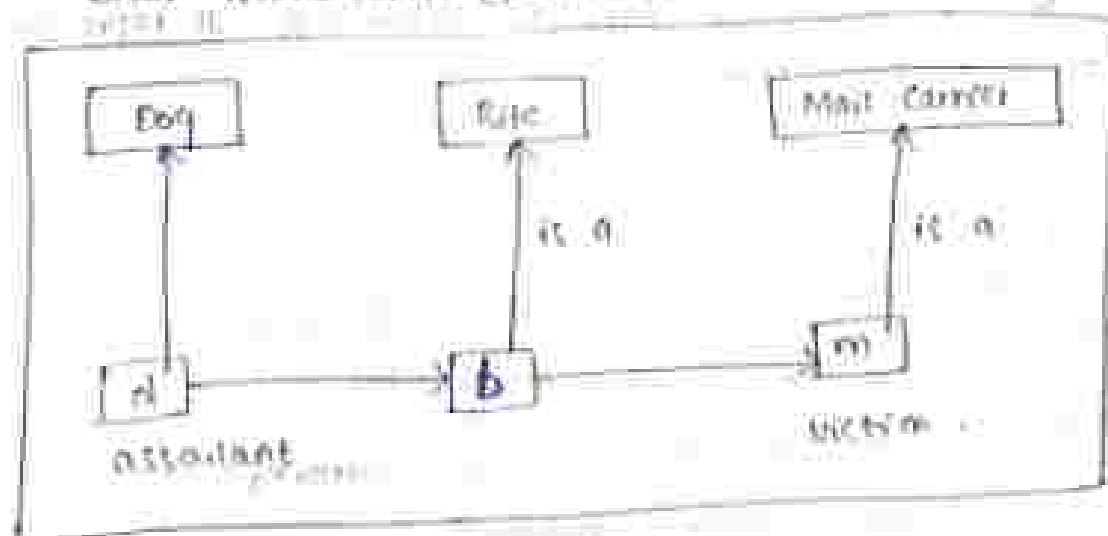
instance

Ex: Create a simple semantic network for a class named person class having properties of class mammal. Person class as instance named Ramu who is in the team xyz having blue colour uniform and person class has a characteristic nose.

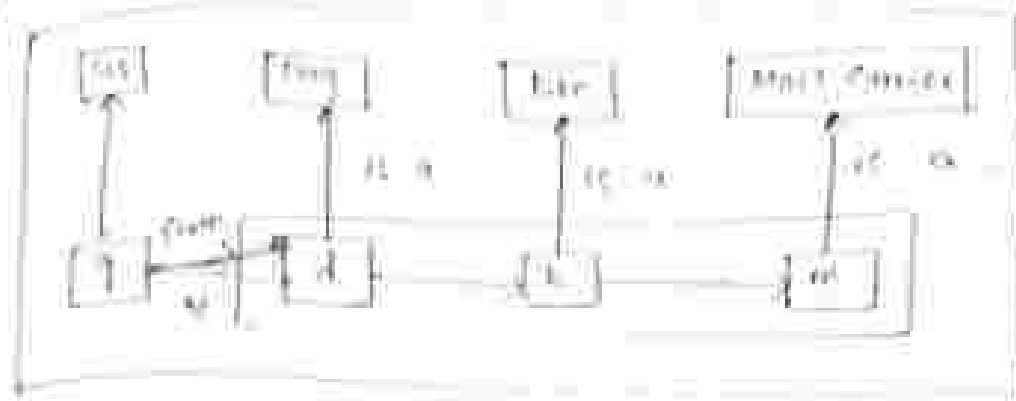


2. partition Semantic network

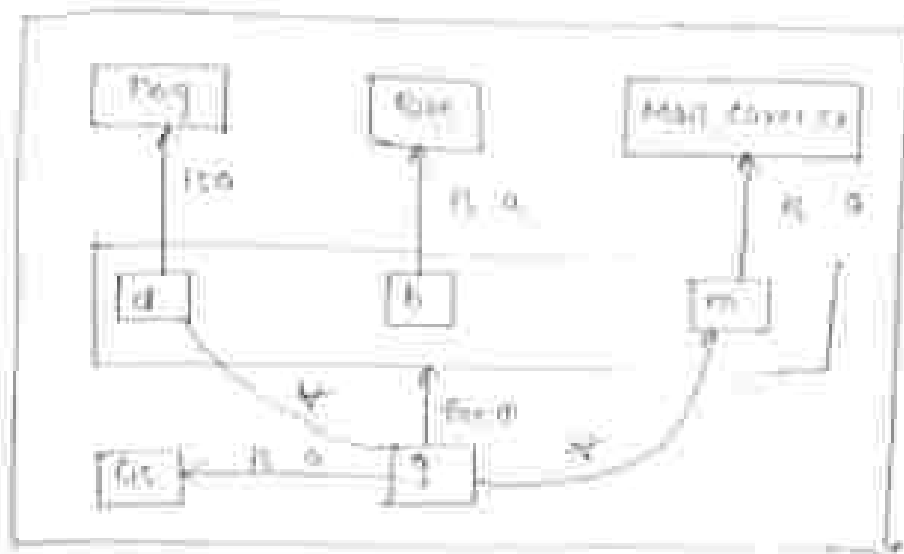
Ex: Draw a partition semantic network for the dog
bird mail-carrier (postman)



every thing has been a node in the frame



every thing in frame has been thing and pointer



Knowledge Representation using Frames:

Frame consists of slots and slots have values. In this system, slots have slots and slots have (associated values) slots and values.

eg. Employee record

In this system, we have slots for slots and slots have values. In this system, we have slots and slots have values.

Slot	Value	Slot	Value
name	John	age	30

For each slot, we have a list of frames:

(slot, (value, slot))

(address (value, bird))

→ Ex: Swamy is yellow bird having wings to fly

Swamy (
 (address (value, bird))
 (color (value, yellow))
 (activity (value, fly))
 (instrument (value, wings))
)

Techniques of Logical Representation:

1. Logical Representation:

In logical representation we have 2 types:

1. Propositional Logic

2. Predicate Logic

Using logical representation we write the simplified form of given statements.

Propositional Logic

Here we use symbols like V, A, N, \rightarrow , \leftrightarrow

Ex: If you score hard then you will achieve everything

This statement is converted by using propositional logic as

score/hard \rightarrow will achieve everything

Predicate Logic

It is an extension to propositional logic

Other than V, A, \rightarrow , \leftrightarrow , N we have quantifiers

as \forall , \exists

Q1: All students like this Sir

$\forall (Student(x) \rightarrow likes(x, Sir))$

2. Semantic Representation:

It is a graphical representation for representing knowledge in patterns of interconnected nodes.

There are two types of Semantic Networks:

1. Simple Semantic Network

2. Partitioned Semantic Network

3. Simple Semantic Network:

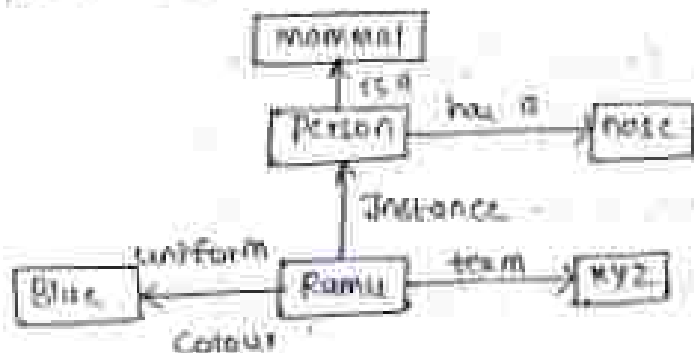
It consists of the following things:

has a

is a

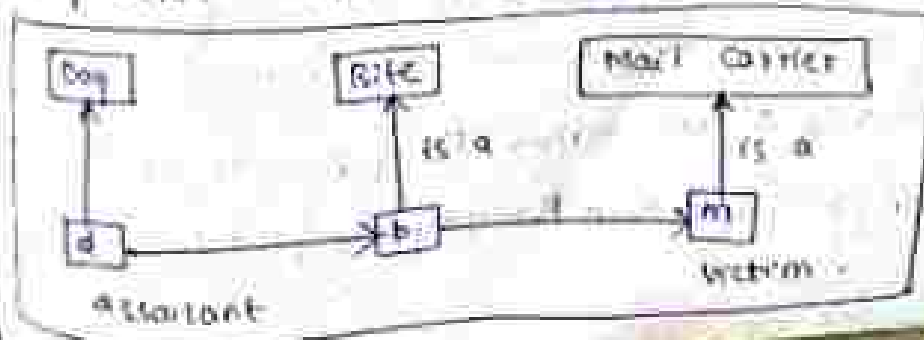
instance

eg: create a simple semantic network for a class
normal person class inherits properties of class
normal. person class as instance named Ramu, who
is in the team xyz having blue colour uniform and
person class has a characteristic role.



4. Partitioned Semantic Network:

Q2: Draw a partitioned semantic network for the
dog bites milk carrier (postman).



3. Frame Representations

These frames are carrying out the records in the database that have both values (attribute values) as in entity:

eg. College details

Its structure will have the table shown below which contains only one record

<u>Roll</u>	<u>Department</u>	<u>Enroll</u>	<u>Address</u>
1001	Management	1010	Delhi

Convert the given statement in terms of frames

(Professor, (value, Manager))

(Enroll, (value, 1010))

(Address, (value, Delhi))

)

Q72

Singgy is yellow bird having wings

Singgy{

(species (value, Bird))

(colour (value, yellow))

(activity (value, fly))

(instrument (value, wings))

}

4. Production Rules

production rules system consist of condition-action pairs which means "If Condition then Action"

Q1: If (bus stop A bus arrives) then (get into the bus)

If (in the bus A paid A empty seat) then (sit down)

if (in bus arrived) then (pay charge)

if (in bus arrived) then (pay charge) else (get down from the bus)

Case Grammar

Case Grammar was proposed by J. Fillmore in 1968.

It is a linguistic (who knows all languages).

He proposed 6 cases:

1. AGENTIVE (Agent)

2. OBJECTIVE (Object)

3. INSTRUMENTAL (Instrument)

4. DATIVE (Indirect Object)

5. FACTIVE (Issue of Action)

6. LOCATIVE (Location of Action)

Table: way to represent cases:

<u>Case</u>	<u>Question</u>
Agent	What was the event occurred?
Agent	Who did the event?
Objective	What was involved in the event?
Beneficiary	Who got benefited with the event?
Time	Which time event was occurred?
Location	Where it was happened?

Example

John gave an apple to Lohari in the kitchen.

Action - Give

Agent - John

Objective - Apple

Beneficiary - Lohari

Time - Past

Location - Kitchen

Here case grammar states that whether the two statements are semantically (or) syntactically equal (or) same.

Eg: Given Statements are:

- (i) The door was broken by John with hammer.
- (ii) Using hammer John broken the door.

Note:

If both the statements i.e. sentences convey the same meaning but may be of different forms then they are called semantically equal.

Eg: (i) Mother baked for one hour.

(ii) Cake baked for one hour.

Note:

In the above statements initial object of the statements is different and rest is same thus these statements are syntactically equivalent (or) equal.

Script:

Event: Student attending exam.

The components provided by the script are:

1. Rules:

These are the persons included in the event
i.e. Student.

2. Props:

Object which are involved in the event.

i.e. pen, pencils, idcard, paper, question paper, etc.

3. Entry condition:

Entry conditions are the conditions needed to be satisfied before an event occurred i.e.
idcard, hallticket.

Object:

Understand that how the exam after exam occurs.

In a other following question paper, he will understand the answers have to markative.

Process:

These are following are variations of the exam is ASSE exam, ETECH exam, GC exam, MBA, BBA's regular etc.

Source:

These are the sequence of actions occurred in exam.

40 Steps:

Exam hall entry



Getting marked seat



Getting question paper



~~Marking~~

Writing answers



Submitting answer sheet

Semantic Web:

The Semantic web is an extension of traditional web. It provides a communication to the user and server allowing the data to share to design web page for use. Hyper text mark up language. But the disadvantage is we cannot use the user defined task. We use only predefined task to that we use Extensible mark up language. It is an extension of HTML.

Eg;

<Circular>

<to> faculty </to>

<from> head </from>

<headline> faculty meeting </headline>

<body> there is meeting of faculty

in room no 17 on

17/10/92 </body>

</Circular>

XML Schema

<Book>

<title>.....</title>

<author>.....</author>

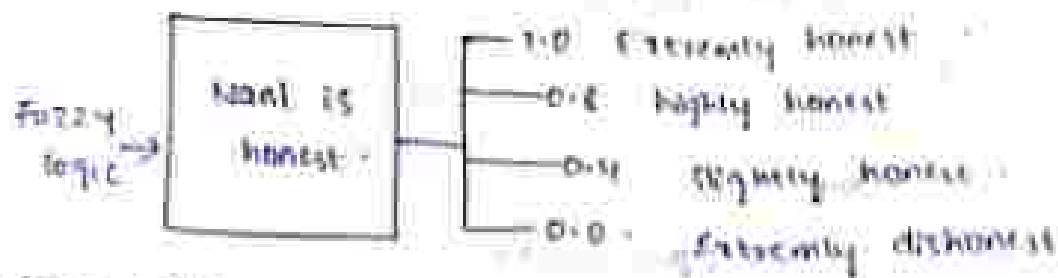
<abstract>.....</abstract>

</book>

Fuzzy Logic:

It is an approach based on degree of truth rather than true or false.

In traditional logic that is predicate and propositional logic, every statement has a truth value that is true or false.



Fuzzy Set:

A regular set is nothing but collection of related items.

Eg: Fruit = {Guava, Apple, banana, fig}

Here Guava, apple, banana, fig are elements of set Fruit.

Fuzzy set:

Fuzzy set is a collection of membership and its type of truth (truth value).

Eg: A = { 0.4 train, 0.6 car, 0.9 cycle } 1.0 = (1)

Membership train slightly belong to set A

Membership car highly belongs to set A

Membership cycle extremely belongs to set A

Linguistic variable

Linguistic variable is a variable which takes words.

$$Ex: Age = \{child, young, old\}$$

The formal definition of a linguistic variable is

$$(X, T(X), U, A(x))$$

where:

X : universe of discourse

$T(X)$: set of terms or values

U : all values of variable for each

A : linguistic rules

$A(x)$: linguistic rules

$$Ex: Age = \{child, young, old\}$$

$$X: age$$

$$T(X): child, young, old$$

U : set of ages of all people

A : give meaning to each term

$$M_{child}(x) : 0 \leq x < 15$$

$$M_{young}(x) : 15 \leq x < 35$$

$$M_{old}(x) : 35 \leq x < 100$$

Fuzzy propositions

P : Ram is a boy

$$T(P) = 0.8$$

\bar{P} : Ram is not a boy

$$T(\bar{P}) = 1 - T(P)$$

$$= 1 - 0.8 \\ = 0.2$$

Fuzzy Proposition is having just a proposition with a truth value.

We can modify given proposition and can combine the given propositions.

Q: Ram is intelligent.

$$T(R) = 0.6$$

And my new proposition is

$P \wedge R$: Ram is boy and Ram is intelligent

$$\therefore T(P \wedge R) = \min\{T(P), T(R)\}$$

$$= \min\{0.4, 0.6\}$$

$$T(P \wedge R) = 0.4$$

Fuzzy Set Operations

The operations performed between Fuzzy sets are

1. Union
2. Intersection
3. Complementary
4. Difference

Union:

If A_1 and A_2 are two fuzzy sets then

$$A_1 \cup A_2 = \max\{\mu_{A_1}(x), \mu_{A_2}(x)\}$$

$$A_1 = \left\{ \frac{1}{10} + \frac{0.75}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\}$$

$$A_2 = \left\{ \frac{1}{10} + \frac{0.6}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0}{3.0} \right\}$$

$$A_1 \cup A_2 = \left\{ \frac{1}{10} + \frac{0.75}{1.5} + \frac{0.3}{2.0} + \frac{0.15}{2.5} + \frac{0}{3.0} \right\}$$

Intersection:

$$A_1 \cap A_2 = \min\{\mu_{A_1}(x), \mu_{A_2}(x)\}$$

$$A_1 \cap A_2 = \left\{ \frac{1}{10} + \frac{0.6}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0}{3.0} \right\}$$

Computation

$$\bar{B}_1 = (1 - B_1(x))$$

Substituting expression from 1

$$\bar{B}_1 = \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0.3}{3.0} \right\}^c$$

$$\bar{B}_1 = (1 - B_1(x))$$

$$\bar{B}_2 = \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.1}{2.0} + \frac{0.2}{2.5} + \frac{0.3}{3.0} \right\}^c$$

Intersection

$$B_1 \mid B_2 = B_1 \cap B_2$$

$$\bar{B}_1 \cap \bar{B}_2$$

$$\left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0.3}{3.0} \right\} \cap$$

$$\left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.1}{2.0} + \frac{0.2}{2.5} + \frac{0.3}{3.0} \right\}$$

= Min. we have to write as it is intersection

$$= \left\{ \frac{0}{1.0} + \frac{0.4}{1.5} + \frac{0.2}{2.0} + \frac{0.1}{2.5} + \frac{0.3}{3.0} \right\}$$

Suppose

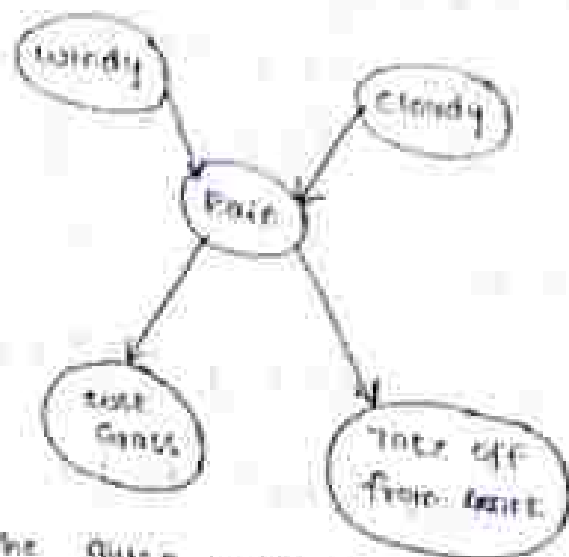
$$B_2 \mid B_1 = B_2 \cap B_1$$

Bayesian belief network:

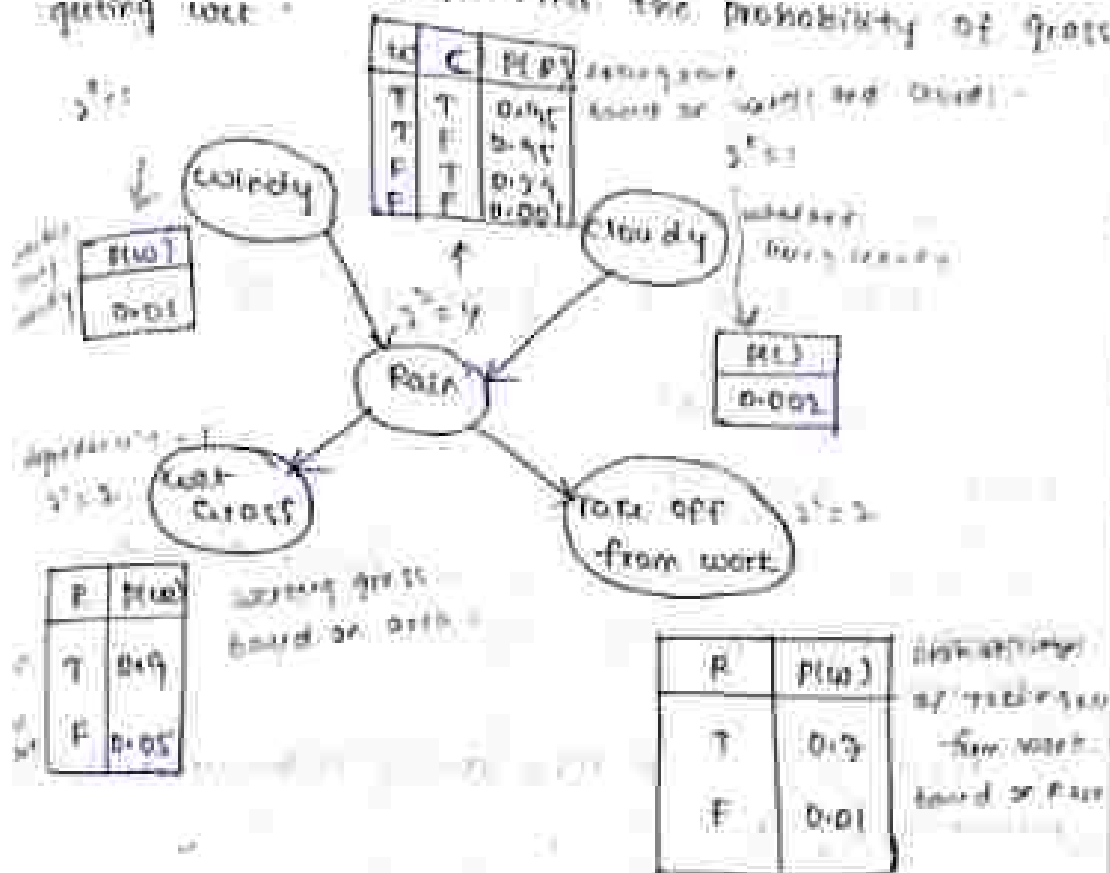
It is a probabilistic graphical model that represents the conditional dependencies between variables through a directed acyclic graph.

$$P(x) = \prod_i P(x_i | \text{parents}(x_i))$$

$$P(x) = \prod_i P(x_i | \text{parents}(x_i)) = \prod_i P(x_i | \text{pa}_i)$$



Q) From the given graph, find the probability of grass getting wet.



Probability of Grass getting wet

$$P(W) = P(W|R) * P(R) + P(W|\bar{R}) * P(\bar{R})$$

$$= 0.9 * P(R) + 0.05 * P(\bar{R})$$

$$P(R) = P(R|W_C) * P(W|C) + P(R|\bar{W}_C) * P(\bar{W}|C)$$

$$+ P(R|W_{\bar{C}}) * P(W|\bar{C}) + P(R|\bar{W}_{\bar{C}}) * P(\bar{W}|\bar{C})$$

$$P(W|C) = P(W|R) * P(R) + P(W|\bar{R}) * P(\bar{R})$$

$$\begin{aligned}
 P(E) &= 0.95 + 0.001 + 0.002 + 0.25 + (1 - 0.001) \times \\
 &\quad 0.002 + 0.95 + 0.991 + (1 - 0.002) \times \\
 &\quad 0.001 + (1 - 0.001) \times (1 - 0.002) \\
 &= 0.00352
 \end{aligned}$$

$$\begin{aligned}
 P(E) &= P(\bar{E} | \omega_1) \times P(\omega_1) + P(E | \bar{\omega}_1) \times P(\bar{\omega}_1) \\
 &\quad + P(E | \omega_2) \times P(\omega_2) + P(\bar{E} | \bar{\omega}_2) \times \\
 &\quad P(\bar{\omega}_2) \\
 &= 0.99244
 \end{aligned}$$

Substituting these in (3) we get

$$\begin{aligned}
 P(\omega) &= (0.9 \times 0.00352) + (0.05 \times 0.99244) \\
 &= 0.0541
 \end{aligned}$$

Dempster-Shafer Theory

In this concept we use a term called Plausibility indicated by "P".

Here we use a term called probability density indicated by "m".

Eg:

Let @ plausibility have the elements
Allergy, flu, cold, pneumonia represented as
below

$$\Theta = \{ \text{Allergy, flu, cold, pneumonia} \}$$

To diagnose for a person to be fever
elements need are flu, cold, pneumonia,
represented by probability density "m".

$m = \{ \text{true, cold, pneumonia} \}$

$m = 0.2$

prob. probability of $m = 1$

probability of $m = 0.2$

but we have not told us the elements of m are 'a'

Complete Bayesian Theory gives the probability of unnecessary things

prob. message = 1 - 0.2

sample = 0.2

→ for the diagnosis of meningitis now

$m = \{ \text{allergy, fever, cold} \}$

$m = 0.2$

message = 0.2

sample = 0.2

→ For the diagnosis of Asthma

$m = \{ \text{allergy, cold, pneumonia} \}$

$m = 0.4$

message = 0.4

sample = 0.4

Certainty Factor:

Certainty factor gives the probability of
fact asserted is true :-

To calculate certainty factor, we have two things

1. measure of belief

2. measure of disbelief

Measure of Error (ME):

$$ME(H, E) = 0$$

H = Hypothesis

E = Evidence

This states that for given evidence, hypothesis is false.

$$ME(H, E) = 1$$

This states that for provided evidence, hypothesis is true.

Measure of Distance (MD):

$$MD(H, E) = 0$$

This states that for the provided evidence, the hypothesis is true.

$$MD(H, E) = 1$$

This states that for the provided evidence, the hypothesis is false.

Multiple evidences Single hypothesis:

$$ME(H, E_1 \text{ and } E_2) = ME(H, E_1) + ME(H, E_2) * [1 - ME(H, E_1)]$$

$$MD(H, E_1 \text{ and } E_2) = MD(H, E_1) + MD(H, E_2) * [1 - MD(H, E_1)]$$

$$CF(H, E_1 \text{ and } E_2) = ME(H, E_1 \text{ and } E_2)$$

$$= MD(H, E_1 \text{ and } E_2)$$

Calculate the certainty factor by knowing the values given.

$$ME(H, E_1) = 0.45, \quad MD(H, E_1) = 0$$

$$ME(H, E_2) = 0.3, \quad MD(H, E_2) = 0.1$$

Probability function

Definition: Probability

$$P(A) = \frac{n(A)}{n(S)} \quad (1)$$

$$P(A) = \frac{n(A)}{n(S)} \quad (2)$$

$$P(A) = \frac{n(A)}{n(S)} \quad (3)$$

$$P(A) = \frac{n(A)}{n(S)}$$

Definition: Probability

$$P(A) = \frac{n(A)}{n(S)}$$

$$P(A) = \frac{n(A)}{n(S)}$$

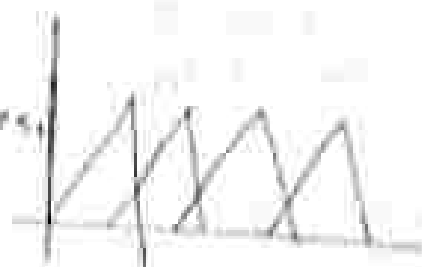
$$P(A) = \frac{n(A)}{n(S)}$$

$$P(A) = \frac{n(A)}{n(S)}$$

Type of membership function

Graph

A fuzzy set consists of membership with truth values, when we plot all these truth values in a graph,



If it represents in the graph drawn above, then we call a fuzzy set of membership type "triangular".

ii)

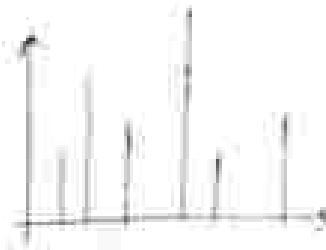
A fuzzy set consists of membership with truth values, when we plot these truth values in a graph, if it represents in the graph drawn above, then



we call a fuzzy set of membership type "trapezoidal".

3. Singleton

A Fuzzy set consists of membership with truth values and when we plot these truth values if we get graph as above called "Singleton".



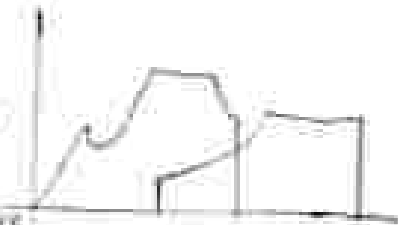
4. Gaussian

A Fuzzy set consists of membership with truth values and when we plot these truth values if we get graph as above it is called "Gaussian".

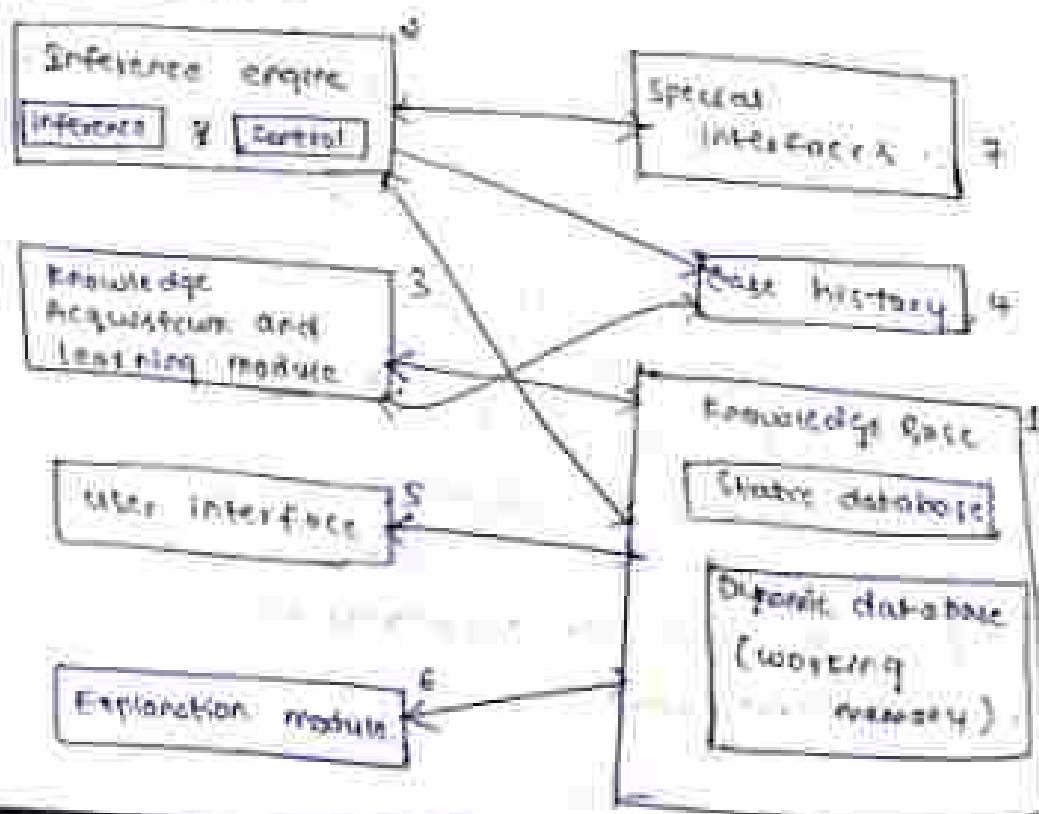


5. Piece wise linear

A fuzzy set consists of membership with truth values and when we plot these truth values if we get graph as above it is called "piece wise linear".



Expert Systems :



Knowledge base:

It is a database of the expert system.

It is of 2 types:

1. Static database

The pre-written AS programs and codes are stored here.

2. Dynamic database:

The intermediate results while solving problem is stored here.

Inference Engine:

It has 2 components:

1. Inference:

Bringing data from secondary to primary memory.

2. Control:

Responsibility of applying the rules to solve the particular problem.

Knowledge Acquisition and Learning Module

Acquisition means addition. To add knowledge to expert system and have to make the system learn.

Partial Solution:

Delivery of the solution without complete information.

Case history:

Statements converted in the form of cases are stored here.

User Interface:

This module helps the communication between machine and human.

Eg: User: hi
Machine: hi

Exploration module

This module explains the user why it came to conclusion.

Eg: Machine: Are you depressed?
User: Why are you asking?
Machine: Your looks were different.
User: Yes.
Machine: By seeing your looks I am concluding that you are depressed.
