

CHAPTER – 7 STATISTICAL REASONING



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Prepared By:

Asst. Prof. Twinkal Panchal

(CSE Department, ACET)

Bayes' Theorem

- Bayes' theorem is also known as **Bayes' rule**, **Bayes' law**, or **Bayesian reasoning**, which determines the probability of an event with uncertain knowledge.
- It is a way to calculate the value of $P(B|A)$ with the knowledge of $P(A|B)$.
- Bayes' theorem can be derived using product rule and conditional probability of event A with known event B:
$$P(A \wedge B) = P(A|B) P(B) \text{ or}$$
- Similarly, the probability of event B with known event A:
$$P(A \wedge B) = P(B|A) P(A)$$
- Equating right hand side of both the equations, we will get:

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)} \quad \dots(a)$$

- The above equation (a) is called as **Bayes' rule** or **Bayes' theorem**. This equation is basic of most modern AI systems for **probabilistic inference**.
- $P(A|B)$ is known as **posterior**, which we need to calculate, and it will be read as Probability of hypothesis A when we have occurred an evidence B.
- $P(B|A)$ is called the likelihood, in which we consider that hypothesis is true, then we calculate the probability of evidence.

- P(A) is called the **prior probability**, probability of hypothesis before considering the evidence
- P(B) is called **marginal probability**, pure probability of an evidence.
- In the equation (a), in general, we can write $P(B) = \sum P(A_i) * P(B|A_i)$, hence the Bayes' rule can be written as:

$$P(A_i | B) = \frac{P(A_i) * P(B|A_i)}{\sum_{i=1}^k P(A_i) * P(B|A_i)}$$

Where $A_1, A_2, A_3, \dots, A_n$ is a set of mutually exclusive and exhaustive events.

Ex-

- A doctor is aware that disease meningitis causes a patient to have a stiff neck, and it occurs 80% of the time. He is also aware of some more facts, which are given as follows:
- The Known probability that a patient has meningitis disease is $1/30,000$. The Known probability that a patient has a stiff neck is 2%.

$$P(a|b) = 0.8$$

$$P(b) = 1/30000$$

$$P(a) = .02$$

$$P(b|a) = \frac{P(a|b)P(b)}{P(a)} = \frac{0.8 * (\frac{1}{30000})}{0.02} = 0.001333333.$$

Application of Bayes' theorem

- It is used to calculate the next step of the robot when the already executed step is given.
- Bayes' theorem is helpful in weather forecasting.
- It can solve the Monty Hall problem.

Certainty factors and rule based system

A certainty factor (CF [h, e]) is defined in terms of two components:

1. MB [h, e]

- A measure (between 0 and 1) of belief in hypothesis h given the evidence e. MB measures the extent to which the evidence supports the hypothesis. It is zero if the evidence fails to support the hypothesis.

2. MD [h, e]

- A measure (between 0 and 1) of disbelief in hypothesis h given the evidence e. MD measures the extent to which the evidence supports the negation of the hypothesis. It is zero if the evidence supports the hypothesis.

$$CF [h, e] = MB[h, e] - MD[h, e]$$

- The measures of belief and disbelief of a hypothesis given, two observations S_1 and S_2 are computed from:

if $MD[h, s_1 \wedge s_2] = 1$ otherwise

$$MB[h, s_1 \wedge s_2] = \begin{cases} 0 \\ MB[h, s_1] + MD[h, s_2] \cdot (1 - MB(h, s_1)) \end{cases} \dots (7.5)$$

if $MB[h, s_1 \wedge s_2] = 1$

$$MD[h, s_1 \wedge s_2] = \begin{cases} 0 \\ MD[h, s_1] + MD[h, s_2] \cdot (1 - MD(h, s_1)) \end{cases} \text{otherwise}$$

Ex -

- Suppose we make an initial observation corresponding to Fig. 7.4 (a) which confirms our belief in h with $MB = 0.3$. Then $MD[h, s_1] = 0$ and $CF[h, s_1] = 0.3$. Now we make a second observation, which also confirms h , with $MB[h, s_2] = 0.2$.

$$\begin{aligned}MH[h, s_1 \wedge s_2] &= 0.3 + 0.2 \times 0.7 \\ &= 0.44\end{aligned}$$

$$MD[h, s_1 \wedge s_2] = 0.0$$

$$CF[h, s_1 \wedge s_2] = 0.44$$

- MYCIN uses for the MB the conjunction and the disjunction of two hypotheses as given below:

$$MB[h_1 \wedge h_2, e] = \min(MB[h_1, e], MB[h_2, e])$$

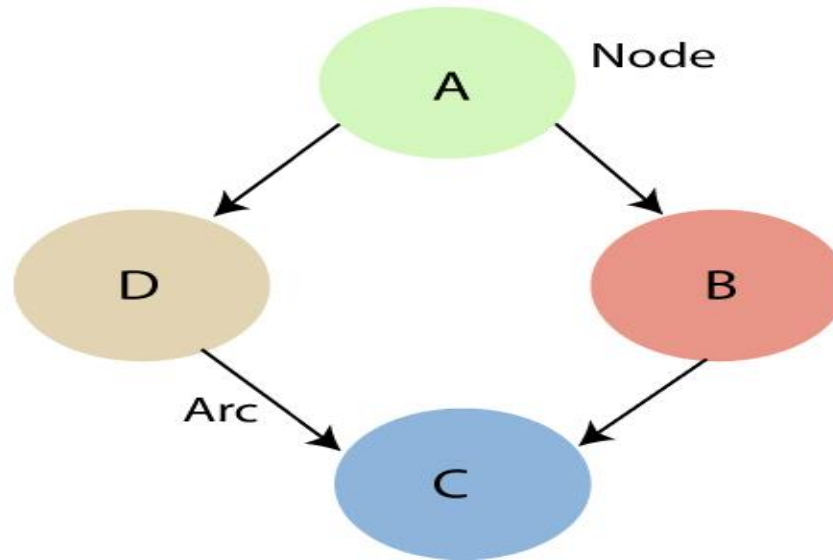
$$MB[h_1 \vee h_2, e] = \max(MB[h_1, e], MB[h_2, e])$$

...(7.6)

Bayesian Network

- “A Bayesian network is a probabilistic graphical model which represents a set of variables and their conditional dependencies using a directed acyclic graph.”
- It is also called a **Bayes network, belief network, decision network, or Bayesian model.**
- Bayesian Network consists of two parts:
 1. Directed Acyclic Graph
 2. Table of conditional probabilities.

- A Bayesian network graph is made up of nodes and Arcs (directed links), where:



- Each node corresponds to the random variables, and a variable can be continuous or discrete.
- Arc or directed arrows represent the causal relationship or conditional probabilities between random variables

- In the above diagram, A, B, C, and D are random variables represented by the nodes of the network graph.
- If we are considering node B, which is connected with node A by a directed arrow, then node A is called the parent of Node B.
- Node C is independent of node A.

Dempster – Shafer Theory

- The Dempster-Shafer theory was designed to mathematically model and validate the uncertainty involved in statistical inferences. This theory allows one to combine evidence from different sources and arrive at a degree of belief, which is mathematically defined by belief functions (Bel).

Belief Functions

- Belief functions were initially proposed as a way to achieve generalized Bayesian inference without priors and mathematically related to random sets. Belief functions combine a set of representations and model data when there is a lack of information.

Belief And Plausibility

- Dempster's rule of combination is sometimes interpreted as an approximate generalisation of Bayes' rule.
- Belief in a hypothesis is calculated as the summation of all these masses of all subsets of the hypothesis-set. The value of belief gives a lower bound on its probability.
- Belief(Bel) ranges from 0 to 1 where 0 indicates no evidence and 1 indicates certainty.

$$\text{Plausibility(Pl)} = 1 - \text{Bel.}$$

Fuzzy Logic

- Fuzzy Logic Systems (FLS) produce acceptable but definite output in response to incomplete, ambiguous, distorted, or inaccurate (fuzzy) input.
- Fuzzy Logic (FL) is a method of reasoning that resembles human reasoning. The approach of FL imitates the way of decision making in humans that involves all intermediate possibilities between digital values YES and NO.
- The conventional logic block that a computer can understand takes precise input and produces a definite output as TRUE or FALSE, which is equivalent to human's YES or NO.

Application Areas of Fuzzy Logic

- 1. Automotive Systems** - Automatic Gearboxes
Four-Wheel Steering
Vehicle environment control
- 2. Consumer Electronic Goods** - Hi-Fi Systems
Photocopiers
Television
- 3. Domestic Goods** - Microwave Ovens
Refrigerators
Toasters
Vacuum Cleaners
Washing Machines

Advantages

- Mathematical concepts within fuzzy reasoning are very simple.
- Fuzzy logic Systems can take imprecise, distorted, noisy input information.
- FLSs are easy to construct and understand.

Disadvantages

- There is no systematic approach to fuzzy system designing.
- They are understandable only when simple.
- They are suitable for the problems which do not need high accuracy.



Thank you!