

**PDPM** 

Indian Institute of Information Technology, Design and Manufacturing, Jabalpur

### Reasoning under Uncertainty

Instructors: Dr. Durgesh Singh

CSE Discipline, PDPM IIITDM, Jabalpur -482005

## Reasoning under uncertainty

- Agents in the real world need to handle uncertainty, whether due to partial observability, nondeterminism, or adversaries.
- An agent may never know for sure what state it is in now or where it will end up after a sequence of actions.

### Nature of Uncertain Knowledge

 Let us try to write rules for dental diagnosis using propositional logic, so that we can see how the logical approach breaks down.
 Consider the following simple rule:

Toothache  $\Rightarrow$  Cavity.

- The problem is that this rule is wrong.
- Not all patients with toothaches have cavities; some of them have gum disease, swelling, or one of several other problems:

Toothache ⇒ Cavity ∨ GumProblem ∨ Swelling ∨ .......

### Nature of Uncertain Knowledge

• In order to make the rule true, we have to add an almost unlimited list of possible problems. We could try turning the rule into a causal rule:

Cavity ⇒ Toothache

But this rule is also not right; not all cavities cause pain. Toothache and a Cavity are always not connected, so the judgement may go wrong.

### Nature of Uncertain Knowledge

- This is typical of the medical domain, as well as most other judgmental domains: law, business, design, automobile repair, gardening, dating, and so on.
- The agent's knowledge can at best provide only a degree of belief in the relevant sentences.
- Our main tool for dealing with degrees of belief is probability theory.
- A logical agent believes each sentence to be true or false or has no opinion, whereas a probabilistic agent may have a numerical degree of belief between 0 (for sentences that are certainly false) and 1 (certainly true).

## **Basic Probability Notation**

- Random variables are typically divided into three kinds, depending on the type of the domain:
- Boolean random variables, such as Cavity, have the domain (true, false) or (1,0)
- Discrete random variables, take on values from a countable domain. For example, the domain of Weather might be (sunny, rainy, cloudy, snow).
- Continuous random variables (bounded or unbounded) take on values from the real numbers. Ex: temp=21.4; temp<21.4 or temp< 1.</li>

## Atomic events or sample points

- Atomic event: A complete specification of the state of the world about which the agent is uncertain
- E.g., if the world consists of only two Boolean variables Cavity and Toothache, then there are 4 distinct atomic events:

```
Cavity = false \wedge Toothache = false
```

Cavity = false  $\wedge$  Toothache = true

Cavity = true  $\wedge$  Toothache = false

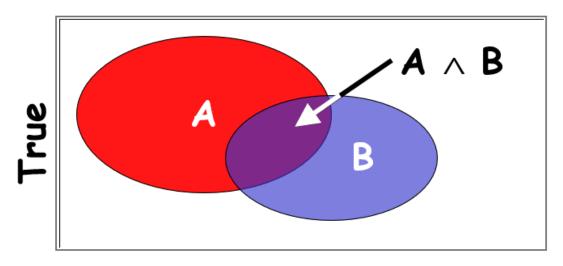
Cavity = true ∧ Toothache = true

- Atomic events are mutually exclusive and exhaustive
- When two events are mutually exclusive, it means they cannot both occur at the same time.
- When two events are exhaustive, it means that one of them must occur.

# **Axioms of Probability Theory**

- All probabilities between 0 and 1
  - $-0 \le P(A) \le 1$
  - -P(true) = 1
  - -P(false) = 0.
- The probability of disjunction is:

$$P(A \lor B) = P(A) + P(B) - P(A \land B)$$



## Prior probability

- The unconditional or prior probability associated with a proposition A is the degree of belief according to the absence of any other information;
- It is written as P (A).
- For example, if the prior probability that I have a cavity is 0.1, then we would write

P (Cavity= true) = 
$$0.1$$
 or P (cavity) =  $0.1$ 

- P (A) can be used only when there is no other information.
- As soon as some new information is known, we must reason with the conditional probability of a given that new information.

## Prior probability...

- Sometimes, we will want to talk about the probabilities of all the possible values of a random variable.
- In that case, we will use an expression such as P(Weather), which denotes a vector of values for the probabilities of each individual state of the weather.
- Instead of writing these four equations

```
P (Weather = sunny) = 0.7
P (Weather= rain) = 0.2
P (Weather= cloudy) = 0.08
P(Weather = snow) = 0.02
```

we may simply write: P(Weather) = (0.7,0.2,0.08,0.02) (Note that the probabilities sum to 1)

This statement defines a prior probability distribution for the random variable Weather.

## Prior probability...

- Joint probability distribution for a set of random variables gives the probability of every atomic event on those random variables
- P(Weather, Cavity) = a 4 × 2 matrix of values:

Weather =	sunny rain	y cloudy snow
Cavity = true	0.144 0.02	0.016 0.02
Cavity = false	0.576 0.08	0.064 0.08

 A full joint distribution specifies the probability of every atomic event and is therefore a complete specification of one's uncertainty about the world in question.

# Conditional or posterior probability

The notation used is P(a | b), where a and b are any proposition.
 This is read as "the probability of a, given that all we know is b."
 For example,

P(cavity I toothache) = 0.8

"indicates that if a patient is observed to have a toothache and no other information is yet available, then the probability of the patient's having a cavity will be 0.8."

# Conditional or posterior probability

 Conditional probabilities can be defined in terms of unconditional probabilities.

$$P(a|b) = \frac{P(a^b)}{P(b)}$$

holds whenever P(b)>0

This equation can be written as

$$P(a^b) = P(a|b) * P(b)$$
 (which is called product rule)

Alternative way:

$$P(a^b) = P(b|a) * P(a)$$

## Chain Rule/Product Rule

$$P(X_1, ..., X_n) = P(X_n | X_1..X_{n-1})P(X_{n-1} | X_1..X_{n-2})... P(X_1)$$
  
=  $\prod P(X_i | X_1,...X_{i-1})$