

# CS 480 – INTRODUCTION TO ARTIFICIAL INTELLIGENCE

TOPIC: UNCERTAINTY  
CHAPTER: 12



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# MEDICAL DIAGNOSIS - TOOTHACHE

- Toothache  $\Rightarrow$  Cavity
  - What's wrong?
    - Not all toothaches are due to cavity
- Toothache  $\Rightarrow$  Cavity  $\vee$  GumProblem  $\vee$  Abscess  $\vee$  ...
  - What's wrong?
    - We have to add almost an unlimited number of possible problems
- Cavity  $\Rightarrow$  Toothache
  - What's wrong?
    - Not all cavities cause toothache

# LOGIC

## ○ Fails for three main reasons

### 1. **Laziness**

- Too much work to list all premises and conclusions
- Too hard to use such rules

### 2. **Theoretical ignorance**

- No complete theory for the domain

### 3. **Practical ignorance**

- Even if we knew all the rules, some information might be missing; e.g., lab tests for diagnosis

# SOME SOURCES OF UNCERTAINTY

## ○ **Uncertainty in knowledge**

- E.g., We do not know all the causes of all the diseases

## ○ **Uncertainty in actions;** we cannot list all the pre-conditions of actions

- E.g., To be able to fly a plane from SFO to JFK, it must not be broken, the weather conditions have to be appropriate, the pilot must not be sick, you need to have enough fuel, ...

## ○ **Uncertainty in sensors**

- E.g., lightning conditions for a camera might not be enough

# TASKS

## 1. Representation

- What is the formal and appropriate language to represent uncertainty?

## 2. Inference

- How can we infer uncertainty before or after we gather more information?

## 3. Decision making

- How can a rational agent act in an uncertain world?

# PROBABILITY MODEL

- It's all about the state the world is in, i.e., a possible world
- A possible world is an assignment of truth values to the predicates
  - Logical assertions rule out some of the possible worlds
    - E.g.,  $\text{cavity} \Rightarrow \text{toothache}$  rules out the worlds where  $\text{cavity}=\text{true} \wedge \text{toothache}=\text{false}$
  - Probabilistic reasoning determines how probable the various worlds are
    - E.g., a world where  $\text{cavity}=\text{true} \wedge \text{toothache}=\text{true}$  is more probable than the world where  $\text{cavity}=\text{true} \wedge \text{toothache}=\text{false}$
- The set of all possible worlds is called the sample space
- The possible worlds are *mutually exclusive* and *exhaustive*
  - *Mutually exclusive*: the world can be only in one state
  - *Exhaustive*: the world has to be in one of the states

# PROBABILITY MODEL

- A **probability model** associates a numerical probability  $P(w)$  with each possible world  $w$ 
  - $P(w)$  sums to 1 over all possible worlds
- An **event** is the set of possible worlds where a given predicate is true
  - Roll two dice; the possible worlds are (1,1), (1,2), ..., (6,6); 36 possible worlds
  - Predicate = two dice sum to 10
  - Event = {(4,6), (5,5), (6,4)}

# AXIOMS OF PROBABILITY

1. The probability  $P(a)$  of a proposition  $a$  is a real number between 0 and 1
2.  $P(\text{true}) = 1$ ,  $P(\text{false}) = 0$
3.  $P(a \vee b) = P(a) + P(b) - P(a \wedge b)$



## $P(\neg a)$

- $P(a \vee \neg a) = P(a) + P(\neg a) - P(a \wedge \neg a)$
- $P(\text{true}) = P(a) + P(\neg a) - P(\text{false})$
- $1 = P(a) + P(\neg a) - 0$
- $P(\neg a) = 1 - P(a)$
- Intuitive explanation:
  - The probability of all possible worlds is 1
  - Either  $a$  or  $\neg a$  holds in one world
  - The worlds that  $a$  holds and the worlds that  $\neg a$  holds are mutually exclusive and exhaustive

# RANDOM VARIABLE

- Like CSP
  - A factored representation of the world; random variables
  - Each variable has a domain
  - Probabilities over the domain values of a variable sum to 1
    - The possible worlds where a random variable takes a certain value are mutually exclusive and exhaustive (from the viewpoint of that variable)
- E.g.
  - $D_1: \{1, 2, 3, 4, 5, 6\}$

# JOINT DISTRIBUTION

- We have  $n$  random variables,  $V_1, V_2, \dots, V_n$
- We are interested in the probability of a possible world, where
  - $V_1=v_1, V_2=v_2, \dots, V_n = v_n$
- $P(V_1, V_2, \dots, V_n)$  associates a probability for each possible world  $\equiv$  the **joint distribution**
  - How many entries are there, if we assume the variables are all binary?
  - How is this related to the truth tables in logic?

# TOOTHACHE EXAMPLE

Toothache	Cavity	P(T,C)
toothache	cavity	0.15
toothache	$\neg$ cavity	0.10
$\neg$ toothache	cavity	0.05
$\neg$ toothache	$\neg$ cavity	0.70

These probabilities are different from what is given in the textbook

# NOTATION

- An upper-case letter  $A_1$  represents a variable and its all possible values
- A lower-case letter  $a_1$  represents a particular value of the variable  $A_1$
- $P(A_1)$  represents a table/function specifying a probability for each possible value of  $A_1$
- $P(A_1=a_1)$  represents a scalar value specifying the probability of  $A_1=a_1$
- We often abbreviate  $P(A_1=a_1)$  as  $P(a_1)$

# PRIOR AND POSTERIOR

- Prior probability
  - Probability of a proposition in the absence of any other information
  - E.g.,  $P(V_1, V_3, V_5)$
- Conditional/posterior probability
  - Probability of a proposition given another piece of information
  - E.g.,  $P(V_2, V_3 \mid V_5 = T, V_7 = F)$
  - $P(A \mid B) = P(A \wedge B) / P(B)$

# NUMBER OF PARAMETERS

- Assuming everything is binary
- $P(V_1)$  requires
  - 1 independent parameter
- $P(V_1, V_2, \dots, V_n)$  requires
  - $2^n - 1$  independent parameters
- $P(V_1 | V_2)$  requires
  - 2 independent parameters
- $P(V_1, V_2, \dots, V_n | V_{n+1}, V_{n+2}, \dots, V_{n+m})$  requires
  - $2^m \times (2^n - 1)$  independent parameters

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# MARGINALIZATION

- Given  $P(V_1, V_2, \dots, V_n \mid V_{n+1}, V_{n+2}, \dots, V_{n+m})$ , where  $n > 0$  and  $m \geq 0$ , we can find, for example
  - $P(V_i, V_j, V_k \mid V_{n+1}, V_{n+2}, \dots, V_{n+m})$  where  $i, j, k < n$  by summing out all the irrelevant variables
- Examples

## LET'S ANSWER A FEW QUERIES

Toothache	Cavity	P(T,C)
toothache	cavity	0.15
toothache	$\neg$ cavity	0.10
$\neg$ toothache	cavity	0.05
$\neg$ toothache	$\neg$ cavity	0.70

- $P(\text{cavity}) = ?$
- $P(\neg \text{cavity}) = ?$
- $P(\text{toothache}) = ?$
- $P(\neg \text{toothache}) = ?$

# LET'S ANSWER A FEW QUERIES

Toothache	Cavity	P(T,C)
toothache	cavity	0.15
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$\neg$ toothache	cavity	0.05
$\neg$ toothache	$\neg$ cavity	0.70

- $P(\text{cavity} \mid \text{toothache}) = ?$
- $P(\text{cavity} \mid \neg \text{toothache}) = ?$
- $P(\neg \text{cavity} \mid \text{toothache}) = ?$
- $P(\neg \text{cavity} \mid \neg \text{toothache}) = ?$
- $P(\text{toothache} \mid \text{cavity}) = ?$
- $P(\neg \text{toothache} \mid \text{cavity}) = ?$
- $P(\text{toothache} \mid \neg \text{cavity}) = ?$
- $P(\neg \text{toothache} \mid \neg \text{cavity}) = ?$

# BAYES' RULE

- $P(B | A) = P(A | B) * P(B) / P(A)$
- Example use
  - $P(\text{cause} | \text{effect}) = P(\text{effect} | \text{cause}) * P(\text{cause}) / P(\text{effect})$
- Why is this useful?
  - Because in practice it is easier to get probabilities for  $P(\text{effect} | \text{cause})$  and  $P(\text{cause})$  than for  $P(\text{cause} | \text{effect})$ 
    - E.g.,  $P(\text{disease} | \text{symptoms}) = P(\text{symptoms} | \text{disease}) * P(\text{disease}) / P(\text{symptoms})$
    - It is easier to know what symptoms diseases cause. It is harder to diagnose a disease given symptoms

# TUBERCULOSIS TEST

## ○ Tuberculosis test

- The test is 90% accurate
  - If you have TB, the test is positive with 90% probability
  - If you don't have TB, the test is negative with 90% probability
- John takes the test and the result is positive
- What is the probability that John has TB?

## ○ Formally

- $P(+ | TB) = 0.9$
- $P(- | \neg TB) = 0.9$
- $P(TB | +) = ?$

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# DECISION MAKING

- Instead of T/F, we have probabilities
- Preferences for certain outcomes (world states)
  - Utility theory
- Maximize expected outcome
  - Decision theory = probability theory + utility theory
- **Maximum Expected Utility** principle
  - An agent is rational if and only if it chooses the action that yields the highest expected utility, averaged over all the possible outcomes of the action