# Topic 4

# Representation and Reasoning with Uncertainty

#### Contents

- 4.0 Representing Uncertainty
- 4.1 Probabilistic methods (Bayesian) PART II
- 4.2 Certainty Factors (CFs)
- 4.3 Dempster-Shafer theory
- 4.4 Fuzzy Logic

#### 4.1 Probabilistic methods (Bayesian)

#### Interpretations of the meaning of "probability"

- **1. Subjective:** the probabilities characterize the beliefs that the observer has about the problem.
- **2. Frequentalist:** the values of the probabilities only can come from repetitions of experiments.

$$p(a) \equiv lim_{\#cases} \rightarrow \infty \left(\frac{\#cases \ of \ a}{\#cases}\right)$$

Practically, we consider:

$$p(a) \approx \frac{\#observed\ cases\ of\ a}{\#observed\ cases}$$

**3. Objective:** the probabilities are real aspects of the Universe, not just beliefs of the observer, or observation.

Interpretation of conditional probability under the frecuency interpretation:

$$p(a \mid b) = \frac{p(a \land b)}{p(b)} =$$

$$= \left(\frac{\#cases\ of\ a \land b}{\#cases}\right) / \left(\frac{\#cases\ of\ b}{\#cases\ of\ b}\right) = \left(\frac{\#cases\ of\ a \land b}{\#cases\ of\ b}\right)$$

In other words, of the cases in which *b* ocurs, the percentage where *a* also occurs.

3

## 4.1 Probabilistic methods (Bayesian)

#### Example:

$$p(a \mid b) = \frac{p(a \land b)}{p(b)} = = \left(\frac{\#cases\ of\ a \land b}{\#cases}\right) / \left(\frac{\#cases\ of\ b}{\#cases}\right) = \left(\frac{\#cases\ of\ a \land b}{\#cases\ of\ b}\right)$$

	has_cold	¬ has_cold
sneezing	15	7
¬ sneezing	5	60

$$p(sneezing|has\_cold) = p(sneezing&has\_cold) / p(has\_cold)$$
$$= 15/92 / (15+5)/92$$
$$= 15 / 20 = 0.75$$

#### **Conditional Probabilities as Rules**

- Conditional Probabilities can be represented as rules:
- p(a|b) = 0.8 if b occurs then a has a 0.8 chance of occuring
- (if (b) then (a)) (0.8)

  if b is true then a is true

  (in 80% of the cases)

5

# 4.1 Probabilistic methods (Bayesian)

## Bayes rule in KBSs

**p** (patient sneezes | patient has cold) = 0.75

Is the same as saying:

**IF** patient has cold **THEN** patient sneezes (0.75)

#### Combining probabilies of premises with rule probabilities

- We have seen that facts can have associated probabilities:
  - p(Pedro has cold) = 0.5
- We have also seen that rules can have associated probabilities (the conditional probability):

**IF** X has cold **THEN** X sneezes (0.75)

- How do we combine these probabilities?
- E.g., if our premises have only partial probability, how probable is our conclusion?

7

#### 4.1 Probabilistic methods (Bayesian)

#### Combining probabilies of premises with rule probabilities

- How do we combine these probabilities?
- The solution is to convert the rule back to a conditional probability format:

**IF** X has cold **THEN** X sneezes (0.75)

- = P(X sneezes | X has cold)=0.75
- We can then produce the probability of the conclusion as follows:
  - We saw earlier that:

$$p(S_1 \wedge S_2) = p(S_2) \cdot p(S_1 \mid S_2)$$

Thus

P(P sneezes & P has cold) = P(P has cold) \* P(P sneezes | P has cold) = 
$$0.5 * 0.75$$
 =  $0.375$ 

#### Another use of the rule: Abduction

IF X has cold THEN X sneezes (0.75)

If the doctor sees Pedro sneezing (absolute certainty). Which is the probability that he has a cold?

We are thus asking:

p (X has a cold | X is sneezing)

when we know...

p (X is sneezing | X has a cold ) = 0.75

How do we calculate  $p(a \mid b)$  from  $p(b \mid a)$ ?

-> Bayes Rule

9

#### 4.1 Probabilistic methods (Bayesian)

#### Another use of the rule: Abduction

Bayes rules:

$$p(x | y) = p(y | x) * p(x) / p(y)$$

We have:

**IF** X has cold **THEN** X sneezing (0.75)

or: p(X sneezing|X has cold) = 0.75

We can calculate:

p(X has cold|X sneezing) = p(X sneezing|X has cold) \*p(X has cold)p(X sneezing)

We thus would need to know:

- p(X has cold) (proportion of the population that has a cold)
- p(X sneezing) (proportion of the population that is sneezing)

#### **Example:**

```
p (sneezing) = 0.4 p (cold) = 0.3 p (sneezing | has cold) = 0.75
```

What is the probability of having a cold if sneezing?

```
p(cold|sneezing) = p(sneezing|cold)*p(cold)/p(sneezing) 
= 0.75 * 0.3 / 0.4 
= 0.56
```

11

#### **Example:**

```
p (sneezing) = 0.1

p (cold) = 0.3

p (sneezing | has cold) = 0.75
```

What is the probability of having a cold if sneezing?

```
Example:

p (sneezing) = 0.1

p (cold) = 0.3

p (sneezing | has cold) = 0.75

What is the probability of having a cold if sneezing?

p(cold|sneezing) = p(sneezing|cold)*p(cold)/p(sneezing)
= 0.75 * 0.3 / 0.1
= 2.25

BUT probabilities have to be between 0 and 1!!!!

Explanation?
```

```
Example:
      p (sneezing) = 0.1
      p \text{ (cold)} = 0.3
      p (sneezing | has cold) = 0.75
What is the probability of having a cold if sneezing?
                            = p(sneezing|cold)*p(cold)/p(sneezing)
      p(cold|sneezing)
                           = 0.75 * 0.3 / 0.1
                            = 2.25
                BUT probabilities have to be between 0 and 1!!!!
It is necessary: p(x) \ge p(x \& y)
We know:
                  p(x | y) = p(x \& y) / p(y)
And thus:
                  p(x \& y) = p(x | y) * p(y)
                  p(sneeze & cold) = p(sneeze|cold) * p(cold)
Substituting:
                                     = 0.75 * 0.3
                                     = 0.225
THUS:
                  p(sneeze) \ge 0.225
```