

# Uncertainty

# Cause of uncertainty

- Most intelligent system may have some degree of uncertainty associated with them
- Reason of uncertainty may be
  - Missing or unavailable data
  - Unreliable or ambiguous data
  - Presence of conflicting measurement techniques
  - knowledge representation which captures knowledge from expert's guess or observation, may not be accurate or correct
- It is not always possible to represent data in precise and consistent manner
- Hence a means to handle uncertainty in the intelligent system is desired

# Uncertainty

- While implementing uncertainty scheme, three issues should be considered
  - Representation of uncertain data
  - Combination of two or more pieces of uncertain data
  - Drawing inference using uncertain data
- Methods available for handling uncertain data is
  - Probability theory
  - Fuzzy logic
  - Temporal logic

# Probability-basic concept

- Probability is likelihood of occurrence of an event or chance that an event will occur.
- $S$  be sample space representing all the possible outcome of an experiment or problem
- Probability of  $A$  is defined as
- $= P(A) = \frac{(\text{number of outcomes favourable to } A)}{\text{Total outcomes}}$
- $= \frac{n(A)}{n(S)}$

# Example

- Consider the experiment of tossing three fair coins, sample space is given by
- $S = \{HHH, HTH, HHT, HTT, TTT, THH, THT, TTH\}$
- $E$  = Event that at least two heads appear  
 $= \{HHH, HTH, HHT, THH\}$
- $F$  = event that first coin shows tail  
 $= \{TTT, TTH, THT, THT\}$   
 $n(E \cap F) = \{TTH\}$
- $P(E) = n(E)/n(S) = 4/8 = 1/2$
- $P(F) = n(F)/n(S) = 4/8 = 1/2$
- $P(E \cap F) = n(E \cap F)/n(S) = 1/8$

# Conditional probability

- Probability of an event E is called conditional probability of E given that event F has already occurred denoted by  $P(E|F)$
- Elements of F which favor E are common elements of E and F ---- sample points  $E \cap F$

$$\begin{aligned}
 P(E|F) &= \frac{n(E \cap F)}{n(F)} \\
 &= \frac{\frac{n(E \cap F)}{n(S)}}{\frac{n(F)}{n(S)}} \\
 &= \frac{P(E \cap F)}{P(F)} \text{ -----(1)}
 \end{aligned}$$

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# Conditional prob

- Example Given that probability of a person chosen at random being literate is 0.40 and probability of any person chosen at random having age  $> 60$  is 0.005. Find Probability that person chosen at random of age  $> 60$  years if literate
- $E$  = event that person age  $> 60$
- $F$  = event that person is literate
- Find  $P(F|E)$
- $P(E)=0.005$      $P(F)= 0.40$
- $P(E \text{ and } F) = P(E) * P(F)= 0.005 * 0.40= 0.002$
- $P(F|E)= P(E \text{ and } F)/ P(E)=0.002/0.005=2/5$

# Joint probability

- Joint probability is defined as the probability of occurrence of two independent events in conjunction
- Probability of both events occurring together
- Joint probability of A and B is written as
  - $P(E \cap F)$  when E and F are dependent

$$P(E \cap F) = P(E|F)P(F) \text{ -----(2)}$$

$$P(E \cap F) = P(E) * P(F) \text{ when E and F Are independent}$$



# $P(E \cup F)$

- $P(E \cup F) = P(E) + P(F) - P(E \cap F)$
- if E and F have some sample points in common
- Otherwise
- $P(E \cup F) = P(E) + P(F)$

# Multiplication theorem of probability

$$P(E \cap F) = P(E | F)P(F) \quad (\text{From equation (2)})$$

Similarly

$$P(E \cap F) = P(F | E)P(E) \quad \text{----- (3)}$$

Using equation 2 and 3

$$P(E | F) * P(F) = P(F | E) * P(E)$$

$$P(E | F) = \frac{P(F | E) * P(E)}{P(F)} \quad \text{----- (4)}$$

This is said to be **Bayes theorem**

# Example

**Example :**An urn contains 10 black and 5 white balls. Two balls are drawn from the urn one after the other without replacement .  
What is the probability that both drawn balls are black

E= First ball drawn is black

F= second ball drawn is black

Find  $P(E \cap F)$

$$P(E) = 10/15$$

$$P(F|E) = 9/14$$

$$P(E \cap F) = (9/14) * (10/15) \quad [Eq(3)]$$

$$= 3/7 \text{ Ans}$$

# Multiplication theorem of probability

- This rule can be extended to more than two events. If E, F and G are the events then
- $P(E \cap F \cap G)$   
 $= P(E) * P(F|E) * P(G|E \cap F)$   
 $= P(E) * P(F|E) * P(G|EF)$

# Joint Distribution

- $k$  random variables  $X_1, \dots, X_k$
- The **joint distribution** of these variables is a table in which each entry gives the probability of one combination of values of  $X_1, \dots, X_k$
- Example:

	Toothache	$\neg$ Toothache
Cavity	0.04	0.06
$\neg$ Cavity	0.01	0.89

$P(\neg\text{Cavity} \wedge \text{Toothache})$

$P(\text{Cavity} \wedge \neg\text{Toothache})$

# Joint Distribution Says It All

	Toothache	$\neg$ Toothache
Cavity	0.04	0.06
$\neg$ Cavity	0.01	0.89

- $P(\text{Toothache}) = P((\text{Toothache} \wedge \text{Cavity}) \vee (\text{Toothache} \wedge \neg \text{Cavity}))$   
 $= P(\text{Toothache} \wedge \text{Cavity}) + P(\text{Toothache} \wedge \neg \text{Cavity})$   
 $= 0.04 + 0.01 = 0.05$
- $P(\text{Toothache} \vee \text{Cavity})$   
 $= P((\text{Toothache} \wedge \text{Cavity}) \vee (\text{Toothache} \wedge \neg \text{Cavity})$   
 $\quad \vee (\neg \text{Toothache} \wedge \text{Cavity}))$   
 $= 0.04 + 0.01 + 0.06 = 0.11$

# Joint Distribution Says It All

	Toothache	$\neg$ Toothache
Cavity	0.04	0.06
$\neg$ Cavity	0.01	0.89

- $P(\text{Toothache}) = ??$
- $P(\text{Toothache} \vee \text{Cavity}) = ??$

# Conditional Probability

- Definition:  
 $P(A|B) = P(A \cap B) / P(B)$
- Read  $P(A|B)$ : probability of A given B
- can also write this as:  
 $P(A \cap B) = P(A|B) P(B)$
- called the **product rule**



# Example

	Toothache	$\neg$ Toothache
Cavity	0.04	0.06
$\neg$ Cavity	0.01	0.89

- $P(\text{Cavity} | \text{Toothache}) = P(\text{Cavity} \wedge \text{Toothache}) / P(\text{Toothache})$ 
  - $P(\text{Cavity} \wedge \text{Toothache}) = ?$
  - $P(\text{Toothache}) = ?$
  - $P(\text{Cavity} | \text{Toothache}) = 0.04/0.05 = 0.8$

# Bayes theorem

- There are two bags I and II
- Bag I contains 2 White and 3 Red balls
- Bag II contains 4 White and 5 Red balls
- We can find
  - Prob of selecting a bag
  - Prob of drawing a ball of a particular color , given the bag
- Can we find
  - Prob that ball is drawn from a particular bag if the color of ball is known?
- This is the case of finding reverse probability using conditional probability
- Formulae developed was given by mathematician Bayes and hence known as bayes theorem

# Theorem of total probability

- Set of events  $E_1, E_2, \dots, E_n$  represents the partition of the sample space  $S$  if

$$(a) E_1 \cap E_2 \cap \dots \cap E_n = \emptyset$$

$$(b) E_1 \cup E_2 \cup \dots \cup E_n = S$$

$$(c) P(E_i) > 0$$

Theorem : Let  $\{E_1, E_2, \dots, E_n\}$  be a partition of the sample space  $S$  and suppose that each of the events  $E_1, E_2, \dots, E_n$  has non zero probability. Let  $A$  be any event associated with  $S$  then

$$\begin{aligned} P(A) &= P(E_1) P(A|E_1) + P(E_2) P(A|E_2), \dots + P(E_n) P(A|E_n) \\ &= \sum_{j=1}^n P(E_j) P(A|E_j) \end{aligned}$$

# Bayes theorem

- If  $E_1, E_2, \dots, E_n$  are non empty events which constitute a partition of sample space  $S$  such that  $E_1 \cup E_2 \cup \dots \cup E_n = S$  and  $E_i$  's are disjoint and  $A$  is any event of non zero probability, then

$$P(E_i | A) = \frac{P(E_i)P(A|E_i)}{\sum_{j=1}^n P(E_j)P(A|E_j)}$$