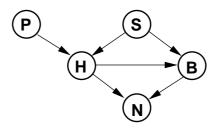
## COMP9414: Artificial Intelligence Solutions Week 6: Reasoning with Uncertainty

- 1.  $P(A \land B) = P(A|B).P(B)$   $P(B \land A) = P(B|A).P(A)$ Now  $P(A \land B) = P(B \land A)$  [why exactly?] Therefore P(A|B).P(B) = P(B|A).P(A)Rearranging gives Bayes' Rule  $P(A|B) = \frac{P(B|A).P(A)}{P(B)}$  if P(B) > 0
- $\begin{array}{l} 2. \ \ P(Mumps|\neg Fever) = \frac{P(\neg Fever|Mumps).P(Mumps)}{P(\neg Fever)} \\ = \frac{(1-P(Fever|Mumps)).P(Mumps)}{1-P(Fever)} \\ = \frac{(1-\frac{3}{4}).\frac{1}{15000}}{1-\frac{1}{1000}} \\ = 0.0000167 \end{array}$
- 3. (i)



$$P(H \land B \land S \land P \land N) = P(H|P \land S).P(B|S \land H).P(S).P(P).P(N|H \land B)$$

(ii) 
$$P(H \land B \land S \land P \land \neg N) = P(H|S \land P).P(B|H \land S).P(S).P(P).P(\neg N|H \land B)$$
  
 $= 0.8 \times 0.4 \times 0.1 \times 0.2 \times 0.1$   
 $= 0.00064$   
 $P(H \land \neg B \land S \land P \land \neg N) = P(H|S \land P).P(\neg B|H \land S).P(S).P(P).P(\neg N|H \land \neg B)$   
 $= 0.8 \times 0.6 \times 0.1 \times 0.2 \times 0.5$   
 $= 0.00480$   
 $P(H \land B \land \neg S \land P \land \neg N) = P(H|\neg S \land P).P(B|H \land \neg S).P(\neg S).P(P).P(\neg N|H \land B)$   
 $= 0.4 \times 0.3 \times 0.9 \times 0.2 \times 0.1$   
 $= 0.00216$   
 $P(H \land \neg B \land \neg S \land P \land \neg N) = P(H|\neg S \land P).P(\neg B|H \land \neg S).P(\neg S).P(P).P(\neg N|H \land \neg B)$ 

$$P(H \land \neg B \land \neg S \land P \land \neg N) = P(H | \neg S \land P).P(\neg B | H \land \neg S).P(\neg S).P(P).P(\neg N | H \land \neg B) = 0.4 \times 0.7 \times 0.9 \times 0.2 \times 0.5$$

= 0.02520

$$P(H \land B \land S \land \neg P \land \neg N) = P(H|S \land \neg P).P(B|H \land S).P(S).P(\neg P).P(\neg N|H \land B)$$
  
= 0.6 \times 0.4 \times 0.1 \times 0.8 \times 0.1

= 0.00192

$$P(H \wedge \neg B \wedge S \wedge \neg P \wedge \neg N) = P(H|S \wedge \neg P).P(\neg B|H \wedge S).P(S).P(\neg P).P(\neg N|H \wedge \neg B) = 0.6 \times 0.6 \times 0.1 \times 0.8 \times 0.5$$

= 0.0144

$$P(H \land B \land \neg S \land \neg P \land \neg N) = P(H | \neg S \land \neg P).P(B | H \land \neg S).P(\neg S).P(\neg P).P(\neg N | H \land B)$$
  
= 0.02 × 0.3 × 0.9 × 0.8 × 0.1

= 0.000432

$$P(H \wedge \neg B \wedge \neg S \wedge \neg P \wedge \neg N) = P(H | \neg S \wedge \neg P).P(\neg B | H \wedge \neg S).P(\neg S).P(\neg P).P(\neg N | H \wedge \neg B)$$

$$=0.02\times0.7\times0.9\times0.8\times0.5$$

= 0.00504

(iii) 
$$\begin{split} P(P|H \land \neg N) &= \frac{P(P \land H \land \neg N)}{P(H \land \neg N)} = \frac{0.0328}{0.054592} = 0.60082 \\ \text{Note:} \\ P(P \land H \land \neg N) &= \Sigma_{b,s} P(H \land b \land s \land P \land \neg N) = 0.00064 + 0.00480 + 0.00216 + 0.02520 \\ P(H \land \neg N) &= \Sigma_{b,s,p} P(H \land b \land s \land p \land \neg N) = 0.00064 + 0.00480 + 0.00216 + 0.02520 + 0.00192 + 0.0144 + 0.000432 + 0.00504 = 0.05452 \end{split}$$

We could also use direct inference, though this provides no advantage in this example. To do this, we need a conditional version of Bayes' Rule:  $P(B|A,C) = \frac{P(A|B,C)P(B|C)}{P(A|C)}$ 

Then 
$$P(P|H, \neg N) = \frac{P(\neg N|H, P).P(P|H)}{P(\neg N|H)}$$

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Now P(\neg N|H, P) = P(\neg N|H, B, P).P(B|H, P) + P(\neg N|H, \neg B, P).P(\neg B|H, P)

= P(\neg N|H, B).P(B|H, P) + P(\neg N|H, \neg B).P(\neg B|H, P)

= P(\neg N|H, B).(P(B|H, S, P).P(S|H, P) + P(B|H, \neg S, P).P(\neg S|H, P)) + P(\neg N|H, \neg B).(P(\neg B|H, S, P).P(S|H, P) + P(\neg B|H, \neg S, P).P(\neg S|H, P))

= [P(\neg N|H, B).(P(B|H, S).P(H|S, P).P(S|P) + P(B|H, \neg S).P(H|\neg S, P).P(\neg S|P)) + P(\neg N|H, \neg B).(P(\neg B|H, S).P(H|S, P).P(S|P) + P(\neg B|H, \neg S).P(H|\neg S, P).P(\neg S|P))]/P(H|P)

= [P(\neg N|H, B).(P(B|H, S).P(H|S, P).P(S) + P(B|H, \neg S).P(H|\neg S, P).P(\neg S)) + P(\neg N|H, \neg B).(P(\neg B|H, S).P(H|S, P).P(S) + P(\neg B|H, \neg S).P(H|\neg S, P).P(\neg S))]/P(H|P)

Also P(P|H) = P(H|P).P(P)/P(H), so cancelling P(H|P)

P(\neg N|H, P).P(P|H)

= [P(\neg N|H, B).(P(B|H, S).P(H|S, P).P(S) + P(B|H, \neg S).P(H|\neg S, P).P(\neg S)) + P(\neg N|H, \neg B).(P(\neg B|H, S).P(H|S, P).P(S) + P(\neg B|H, \neg S).P(H|\neg S, P).P(\neg S))].P(P)/P(H)
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This generates four terms exactly as above, and similarly  $P(\neg N|H)$  generates eight terms as above. The extra P(H) cancel each other out.

4. Let A stand for "Alarm", B for "Burglary" and E for "Earthquake".

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Then by Bayes' Rule
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 $P(B|A) = P(A|B).P(B)/P(A) = (P(A|B \land E).P(E).P(B) + P(A|B \land \neg E).P(\neg E).P(B))/P(A)$ , and as in lectures

$$P(A) = P(A|B \wedge E).P(E).P(B) + P(A|B \wedge \neg E).P(\neg E).P(B) + P(A|\neg B \wedge E).P(E).P(\neg B) + P(A|\neg B \wedge \neg E).P(\neg E).P(\neg B)$$

So 
$$P(B|A) = (0.95 \times 0.002 \times 0.001 + 0.94 \times 0.998 \times 0.001)/P(A)$$
 and  $P(A) = 0.95 \times 0.002 \times 0.001 + 0.94 \times 0.998 \times 0.001 + 0.29 \times 0.002 \times 0.999 + 0.001 \times 0.998 \times 0.999$ 

Thus P(B|A) = 0.00094002/0.002516442 = 0.3735512

Intuitively, the "true positives" (when there really is a burglary) account for roughly only 10/26 of the cases when the alarm is ringing (around 0.001 of the time), while the "false positives" account for 16/26 cases (6/26 when the alarm is ringing because of an earthquake, due to a false positive rate around 0.3 and prior of 0.002, so around 0.0006 of the time, and 10/26 when there is neither a burglary nor an earthquake, due to a false positive rate of 0.001 and a prior close to 1, so around 0.001 of the time). The rough calculation is 10/26 = 0.001/(0.001 + 0.0006 + 0.001). That is, the false positives significantly outweigh the true positives in this scenario.

5. By the Chain Rule,  $P(A \land B \land C) = P(B|A, C).P(A|C).P(C) = P(A|B, C).P(B|C).P(C)$ . So, provided  $P(C) \neq 0$  and  $P(A|C) \neq 0$ , the conditional version of Bayes' Rule follows.