

## Conditioning and Bayes' Rule

(Always Expect the unexpected)

① prob  $\rightarrow$  not impossible  
 $\hookrightarrow$  unlikely to happen.

$P(A|B)$  = probability of A,  
 given that B occurred.

$\hookrightarrow$  B's new universe.

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \text{Assuming } P(B) \neq 0$$

$$P(B|B) = \frac{P(B \cap B)}{P(B)} = \frac{P(B)}{P(B)} = 1$$

$$P(A \cap B) = P(B) \times P(A|B) \\ = P(A) \times P(B|A)$$

$$\text{If } A \cap B = \emptyset$$

$$P(A \cup B) = P(A) + P(B)$$

C occurred

$$P(A \cup B | C) = P(A|C) + P(B|C)$$

examp. 6

Event A  $\rightarrow$  Airplane is flying above

Event B  $\rightarrow$  Something registers on Radar screen

Multiplication

$$P(A \cap B | C) = P(A) \times P(B|A) \times P(C|A \cap B)$$



proof

$$p(A \cap B \cap C) = p((A \cap B) \cap C)$$

$$p(A \cap B \cap C) = p(A \cap B) \times p(C | A \cap B) \\ = p(A) \times p(B | A) \times p(C | A \cap B)$$

Total prob theorem

→ Divide and conquer

→ Partition of sample space into  $A_1, A_2, A_3$

→ Have  $p(B | A_i)$  for every  $i$

$$p(B) = p(A_1) \times p(B | A_1) \\ + p(A_2) \times p(B | A_2) \\ + p(A_3) \times p(B | A_3)$$

Given

$$p(A_1) + p(A_2) + p(A_3) = 1$$

basically weighted averages

## Bayes's Rule

We know  $p(B | A_i)$  has for each  $i$ .

wish to reverse  $p(A_i | B)$

$$p(A_i | B) = \frac{p(A_i \cap B)}{p(B)} = \frac{p(A_i) \times p(B | A_i)}{p(B)} \\ = \frac{p(A_i) \times p(B | A_i)}{\sum_j p(A_j) \times p(B | A_j)}$$

(Causal Effect Model)

