

## Contents

|                         |   |
|-------------------------|---|
| Conditional Probability | 1 |
| Marginalisation         | 2 |
| Bayes Rule              | 2 |

## Conditional Probability

**Conditional probability** is the probability that event  $E$  occurs *given* that event  $F$  has already occurred. Call this “conditioning” on  $F$ . Written as  $P(E | F)$ .

- Same meaning as  $P(E \text{ given } F \text{ already observed})$
- Its a probability (satisfies all the axioms, will see this shortly), with:
  - Sample space  $S$  restricted to those outcomes consistent with  $F$ , i.e.  $S \cap F$
  - Event space  $E$  restricted to those outcomes consistent with  $F$ , i.e.  $E \cap F$

With equally likely outcomes

$$P(E | F) = \frac{\text{number of outcomes in } E \text{ consistent with } F}{\text{number of outcomes in } S \text{ consistent with } F} = \frac{|E \cap F|}{|S \cap F|}$$

Note that  $|S \cap F| = |F|$  so

$$P(E | F) = \frac{|E \cap F|}{|S \cap F|} = \frac{|E \cap F|}{|F|} = \frac{|E \cap F|}{|S|} \frac{|S|}{|F|} = \frac{\frac{|E \cap F|}{|S|}}{\frac{|F|}{|S|}} = \frac{P(E \cap F)}{P(F)}$$

General definition:  $P(E | F) = \frac{P(E \cap F)}{P(F)}$  where  $P(F) > 0$ .

Implies  $P(E \cap F) = P(E | F)P(F)$  known as the *chain rule* - it's important!

If  $P(F) = 0$ ?

- $P(E | F)$  is undefined
- Can't condition on something that can't happen

$P(E|F)$  is a probability - it satisfies all the properties of ordinary probabilities

- $0 \leq P(E | F) \leq 1$
- $P(S | F) = 1$
- If  $E_1, E_2$  are mutually exclusive events then  $P(E_1 \cup E_2 | F) = P(E_1 | F) + P(E_2 | F)$

## Marginalisation

Suppose we have mutually exclusive events  $F_1, F_2, \dots, F_n$  such that  $F_1 \cup F_2 \cup \dots \cup F_n = S$  then

$$P(E) = P(E \cap F_1) + P(E \cap F_2) + \dots + P(E \cap F_n)$$

Marginalisation is very handy, example:

- Roll two coins. What is the probability that the first coin is heads?
- Event E is the first coin heads,  $F_1$  is second coin heads,  $F_2$  is second coin tails
- $P(E) = P(E \cap F_1) + P(E \cap F_2) = \frac{1}{2} \times \frac{1}{2} + \frac{1}{2} \times \frac{1}{2} = \frac{1}{2}$

## Bayes Rule

Recall  $P(E \cap F) = P(E | F)P(F)$

Clearly, and also  $P(F \cap E) = P(F | E)P(E)$

But  $P(E \cap F) = P(F \cap E)$ , so  $P(E | F)P(F) = P(F | E)P(E)$

i.e.  $P(E | F) = \frac{P(F|E)P(E)}{P(F)}$