

# Lecture 6: Uncertainty and Information

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## 1 Introduction

### Sources of Uncertainty

**Complexity** : the number of possible states of a system in question can be too large (e.g. predict how a chess game can develop after 10 moves?)

**Ignorance** : some important information about the system may not be available

**Randomness** : the system may be random by nature, and thus the uncertainty is irreducible.

### Historical Background

**1654** Blaise Pascal and Pierre Fermat

**1657** Christian Huygens publishes *On Ratiocination in Dice Games*

**1760** Thomas Bayes (conditional probability)

**1812** Pierre-Simon Laplace

**1933** Andrey Kolmogorov's axioms

**1920–1940** Ronald Fisher, Abraham Wald (statistics)

**1948** Claude Shannon (Information theory)

## 2 What is probability?

### What is Probability?

**Definition 1.** The uncertainty about some event  $E$  can range from impossible to certain. Let us denote the *probability* of event  $E$  by  $P(E)$  such that

$P(E) = 0$  means  $E$  is impossible;

$P(E) = 1$  means  $E$  is certain.

Thus, probability is a number between 0 and 1.

$$\text{(Impossible)} \quad 0 \leq P(E) \leq 1 \quad \text{(Certain)}$$

*Example 2.* For a fair coin,  $P(\text{heads}) = \frac{1}{2} = 0.5$

### Set-Theoretic Intuition

- Events  $E$  are considered as subsets of the universal set  $U$ :

$$E \subseteq U$$

- Probability is a *measure* of a subset  $E \subseteq U$  such that

$$P(U) = 1$$

(because the universe is certain).

- We can consider probabilities of negation, disjunction and conjunction of events:

$$P(\text{not } E), \quad P(E_1 \text{ or } E_2), \quad P(E_1 \text{ and } E_2)$$

- This corresponds to probability measure of the complement, union and intersection of the subsets:

$$P(\bar{E}) = P(U - E), \quad P(E_1 \cup E_2), \quad P(E_1 \cap E_2)$$

### Additivity of Probabilities

- Events  $E_1$  and  $E_2$  are *disjoint* if  $E_1 \cap E_2 = \emptyset$ .

- For disjoint events

$$P(E_1 \text{ or } E_2) = P(E_1) + P(E_2)$$

- For  $n$  disjoint events such that  $E_1 \cup E_2 \cup \dots \cup E_n = U$

$$P(E_1) + P(E_2) + \dots + P(E_n) = 1$$

(because at least one of the events is certain)

*Example 3.* For a fair coin and a fair dice we have

$$\frac{1}{2} + \frac{1}{2} = 1 \qquad \frac{1}{6} + \frac{1}{6} + \frac{1}{6} + \frac{1}{6} + \frac{1}{6} + \frac{1}{6} = 1$$



### Probability of Negation

- Negation of an event ‘not  $E$ ’ refers to its complement

$$\bar{E} = U - E$$

- $E$  and  $\bar{E}$  are disjoint and such that

$$E \cup \bar{E} = E \cup (U - E) = U$$

- Because  $P(E \text{ or not } E) = P(E) + P(\text{not } E) = 1$  we have

$$P(\text{not } E) = 1 - P(E)$$

**Question 1.** *What is the probability of a void set  $P(\emptyset)$ ?*

### Joint Probability

- Co-occurrence of events  $E_1$  and  $E_2$  (e.g. clouds and rain) is their set intersection:

$$E_1 \cap E_2$$

- Probability of  $E_1 \cap E_2$  is called *joint* probability and denoted  $P(E_1 \cap E_2)$  or simply  $P(E_1, E_2)$ .

- For disjoint events  $P(E_1 \cap E_2) = P(\emptyset) = 0$ .

*Example 4* (Two coins).

$E_1$	$E_2$
heads	heads
heads	tails
tails	heads
tails	tails

### Where do Probabilities Come From?

- If there are  $n$  disjoint events, then we could assume that all

$$P(E_1) = P(E_2) = \dots = P(E_n) = \frac{1}{n}$$

- It would be much better to use the *empirical frequency* function

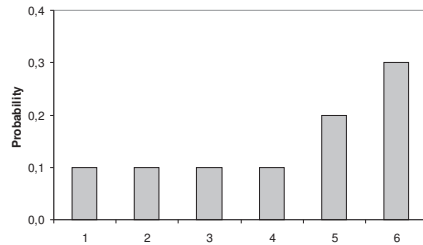
$$P(E_i) \approx \frac{n(E_i)}{n} = \frac{\text{no. of times event } E_i \text{ occurs}}{\text{no. of independent tests}}$$



*Example 5.* Flip a coin or roll a dice several times to estimate the probabilities.

### Probability Distributions

We can plot probabilities of all events on a graph, which shows probability distribution



## 3 Conditional Probability and Independence

### Conditional Probability

**Question 2.** *How likely is it to rain if you see clouds?*

- Probability of  $E_1$  conditional to event  $E_2$  (i.e. if  $E_2$  has happened):

$$P(E_1 | E_2)$$

- Joint probability  $P(E_1 \cap E_2)$  can be computed using conditional by multiplication:

$$P(E_1 \cap E_2) = P(E_1 | E_2)P(E_2)$$

$$P(E_1 \cap E_2) = P(E_2 | E_1)P(E_1)$$

- If  $P(E_2) > 0$ , then the *Bayes formula* holds

$$P(E_1 | E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)}$$

### Independence

**Question 3.** • *How likely is it to rain?*

- *How likely is it to rain if you see clouds?*
- *Are rain and clouds independent?*

**Definition 6.** Event  $E_1$  is *independent* of  $E_2$  if

$$P(E_1 | E_2) = P(E_1)$$

(i.e. knowledge of  $E_2$  does not change the probability of  $E_1$ )

Because  $P(E_1 \cap E_2) = P(E_1 | E_2)P(E_2)$ , the joint probability of independent events is

$$P(E_1 \cap E_2) = P(E_1)P(E_2)$$

### A Simple Credit Score Example

- You selected at random records of 20 customers and divided them based on *homeowner* and *credit score*.

		Credit score	
		Low	High
Homeowner	No	7	3
	Yes	2	8

- Can you tell from this data whether credit score and homeownership depend on each other?

### A Simple Credit Score Example (sol.)

		Credit score		
		Low	High	
Homeowner	No	7	3	10
	Yes	2	8	10
		9	11	20

$$P(H, C) = \left\{ \frac{7}{20}, \frac{2}{20}, \frac{3}{20}, \frac{8}{20} \right\}$$

$$P(H) = \left\{ \frac{10}{20}, \frac{10}{20} \right\}, \quad P(C) = \left\{ \frac{9}{20}, \frac{11}{20} \right\}$$

Our simple test for independence  $P(H, C) \stackrel{?}{=} P(H)P(C)$

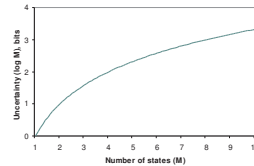
## 4 Uncertainty and Information

### Uncertainty and Information

- If each event can have 2 states (i.e. True, False), then for a system that consists of  $H$  number of events there are

$$M = 2^H \text{ possible states}$$

- To measure uncertainty we can use  $H = \log_2 M$  (aka *entropy*)



- Information reduces uncertainty

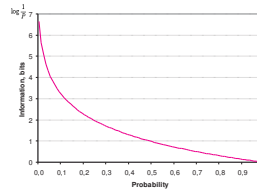
$$\text{Information} = \text{Uncertainty before} - \text{Uncertainty after}$$

- Information can be seen as negative entropy  $I = -H$

### Information and Probability

- What if probabilities are not equal? (i.e.  $P(E) \neq \frac{1}{M}$ )
- We can express the uncertainty as

$$H(E) = \log_2 \frac{1}{P(E)} = -\log_2 P(E)$$



*Example 7.* Compare information from observing events with probabilities  $\frac{1}{8}$  and  $\frac{1}{2}$