

## Combinatorics and graph theory

Lecture note 13. Probability - Finite probability spaces

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We will see a remarkable mathematical application of probability, namely how mathematical statement can be proved using elementary probability theory.

**Definition 1.** *By a finite probability space, we understand a pair  $(\Omega, P)$ , where  $\Omega$  is a finite set and  $P : 2^\Omega \rightarrow [0, 1]$  is a function assigning a number from the interval  $[0, 1]$  to every subset of  $\Omega$ , such that*

- $P(\emptyset) = 0$ ,
- $P(\Omega) = 1$ ,
- $P(A \cup B) = P(A) + P(B)$  for any two disjoint sets  $A$  and  $B$  in  $\Omega$ .

The elements of  $\Omega$  are called the *elementary events*.

For instance, if the experiment is rolling one fair die, the elementary events would be ‘1 was rolled’, ‘2 was rolled’ and so on. We can denote by  $w_1, w_2, \dots$ ,

Subsets of  $\Omega$  are called *events*. For  $A \subseteq \Omega$ ,  $P(A)$  is the probability of the event  $A$ .

Several important species of finite probability spaces.

**Definition 2** (A random sequence of  $n$  0s and 1s). *The elementary events of this probability space are all  $n$ -term sequences of 0s and 1s, that is, elements of the set  $\{0, 1\}^n$ , and all elementary events have the same probability. The probability of each event  $A$  equals  $|A|/2^n$ . We denote this by  $\mathcal{C}_n$ .*

This probability space models a sequence of  $n$  coin tosses, for instance. An example of an event is  $A$ : “Heads appear exactly 10 times”, whose prob is  $\binom{n}{10}/2^n$ .

**Definition 3** (A random permutation). *Elementary events of this probability space are all permutations of the set  $\{1, 2, \dots, n\}$ , and the prob of an event  $A$  is equal  $|A|/n!$ . The set of all permutations is denoted by  $S_n$ , and this probability space by  $\mathcal{S}_n$ .*

PROBLEM : What is the probability that in a random ordering of  $\{1, 2, \dots, 6\}$ , 1 precedes 2?

It can be formulated as the event  $A = \{\pi \in S_5 : \pi(1) < \pi(2)\}$ .

**Definition 4** (A random graph). *This probability space, denoted by  $\mathcal{G}_n$ , has all the possible graphs on vertex set  $\{1, 2, \dots, n\}$  as elementary events, and all of them have the same probability equal to  $2^{-\binom{n}{2}}$ .*

*Examples of events :*

- $S =$  “the graph  $G$  is connected ”
- $B =$  “ the graph  $G$  is bipartite ”

Comparing such events is often very difficult, but we are usually interested in a very rough estimate. For two mentioned examples, it turns out that for  $n \rightarrow \infty$ ,  $P(S)$  rapidly tends to 1 while  $P(B)$  approaches to 0 very quickly.

This is sometimes expressed as

- a random graph is almost surely connected,
- almost surely, a random graph is not bipartite.

**Proposition 1.** *A random graph almost surely is not bipartite. That is  $\lim_{n \rightarrow \infty} P(B) = 0$ .*

*PROOF*

**Independent events.** Two events  $A, B$  in a probability space are called *independent* if we have

$$P(A \cap B) = P(A)P(B).$$

Independence means that if  $\Omega$  is divided into two parts,  $A$  and its complement, the event  $B$  cuts both these parts in the same ratio.

Most often, we encounter independent events in the following situation: The elements of  $\Omega$  can be viewed as ordered pairs, so  $\Omega$  models the possible results of an experiment consisting of two experiments, where result of the first experiment cannot influence the result of the second experiment and vice versa.

The space  $\mathcal{C}_n$  is a typical source of such situations. Similarly, ‘the graph  $G$  has at least one triangle on  $\{1, 2, \dots, 10\}$ ’ and ‘the graph  $G$  contains a cycle on  $\{11, 12, \dots, 20\}$ ’ are independent.

**Definition 5.** Events  $A_1, A_2, \dots, A_n \subseteq \Omega$  are independent if for each set of indices  $I \subseteq \{1, 2, \dots, n\}$ ,

$$P\left(\bigcup_{i \in I} A_i\right) = \prod_{i \in I} P(A_i).$$