- 1. Reading: K. Rosen Discrete Mathematics and Its Applications, 4.5
- 2. The main message of this lecture:

## Probability with not necessarily equally likely outcomes, conditional probability, independent events: all have natural and extremely useful mathematical definitions.

The classical definition of probability (Laplace) assumes that the sample space is finite  $S = \{s_1, s_2, \ldots, s_n\}$ , that all the outcomes  $s_i$  are equally likely and introduces the formula for the probability of an event E:

$$p(E) = \frac{|E|}{|S|}.$$

We can present the same formula in a somewhat more natural way, via the probabilities of individual outcomes. Note that the probability of each single outcome  $p_i = p(s_i) = 1/n$ , and that  $p_1 + p_2 + \ldots + p_n = 1$ . Then p(E) equals to the sum of those p(s) for which  $s \in E$ :

$$p(E) = \sum_{s \in E} p(s) = |E| \cdot \frac{1}{n} = \frac{|E|}{|S|}.$$

**Definition 20.1.** Imagine that S is still finite  $S = \{s_1, s_2, \ldots, s_n\}$ , but the outcomes  $s_i$  are not necessarily equally likely. We assume that the probability of individual outcomes  $p_i = p(s_i)$  are given and that

1.  $0 \le p(s) \le 1$  for each  $s \in S$ 

2.  $p(s_1) + p(s_2) + \ldots + p(s_n) = 1$ .

Then the formula number two for the probability of an event E still applies:

$$p(E) = \sum_{s \in E} p(s).$$

**Example 20.2.** Biased coin: heads come up twice as often as tails.  $S = \{H, T\}, p(H) + p(T) = 1, p(H) = 2p(T), 2p(T) + p(T) = 1$ , therefore, 3p(T) = 1, p(T) = 1/3, p(H) = 2/3.

**Theorem 20.3.**  $p(E_1 \cup E_2) = p(E_1) + p(E_2) - p(E_1 \cap E_2)$ **Proof.** Similar to the Inclusion-Exclusion Principle, in

$$p(E_1) + p(E_2) = \sum_{s \in E_1} p(s) + \sum_{s \in E_2} p(s)$$

each element of the intersection  $E_1 \cap E_2$  is counted twice. Subtracting  $p(E_1 \cap E_2)$  to compensate this overcount we get the desired formula for  $p(E_1 \cup E_2)$ .

**Corollary 20.4.**  $p(\overline{E}) = 1 - p(E)$ . Indeed, since p(S) = 1, by 20.3, we have  $1 = p(S) = p(E \cup \overline{E}) = p(E) + p(\overline{E})$ . Thus  $p(\overline{E}) + p(E) = 1$  and  $p(\overline{E}) = 1 - p(E)$ .

**Example 20.5.** Flipping a fair coin the probability of having at least one T (event E) is 7/8. Suppose we know that the first flip came up heads (event  $F = \{HTT, HTH, HHT, HHH\}$ ) and we want to evaluate the "new" probability of E given F. A new provisional sample set is

now down to F, we have four equally likely outcomes. Some of the outcomes from E (namely THH, TTH, THT, TTT) are no longer possible. To evaluate the "conditional" probability of E given F we have to take the ratio of what is left of E to what is left of S:

$$\frac{|E \cap F|}{|F|} = \frac{|E \cap F|/|S|}{|F|/|S|} = \frac{p(E \cap F)}{p(F)} = \frac{3/8}{4/8} = 3/4.$$

**Definition 20.6. Conditional probability** p(E|F) ("the probability of E given F") is

$$p(E|F) = \frac{p(E \cap F)}{p(F)}.$$

The reason for this definition is similar to the one used in example 20.5: given a condition F we may regard it as a new sample space and adjust the formula accordingly.

**Example 20.7.** What is the conditional probability that a family of three children has more than one boy given they have at least one boy.  $E = \{BBB, BBG, BGB, GBB\}, F = \{BBB, BBG, BGB, GBB, BGG, GBG, GGB\}$ . Then  $E \cap F = E$ , since  $E \subset F$ .

$$p(E|F) = \frac{p(E \cap F)}{p(F)} = \frac{p(E)}{p(F)} = \frac{4/8}{7/8} = 4/7.$$

What if the condition F is "the first child is a girl"? Then  $F = \{GBB, GBG, GGB, GGG\}, E \cap F = \{GBB\}, p(E|F) = p(E \cap F)/p(F) = (1/8)/(4/8) = 1/4.$ 

**Definition 20.8.** If a condition F does not change the probability of an event E, we say that E and F are **independent**: p(E) = p(E|F). Note that in a full accord to intuition, independence is symmetric: E is independent from F if and only if F is independent from E:

$$p(E) = \frac{p(E \cap F)}{p(F)} \quad \Leftrightarrow \quad p(E) \cdot p(F) = p(E \cap F) \quad \Leftrightarrow \quad p(F) = \frac{p(E \cap F)}{p(E)},$$

therefore,  $p(E) = p(E|F) \iff p(F) = p(F|E)$ .

**Example 20.9.** A fair coin is flipped twice. E="the first flip come up tails" ={TT, TH}, F="the second flip comes up tails" ={TT, HT}. Intuitively, E and F are independent. Let us check the formula. On one hand, p(E) = 1/2. On the other hand,  $p(E \cap F) = p({TT}) = 1/4$ ,  $p(E|F) = p(E \cap F)/p(F) = (1/4)/(1/2) = 1/2$ . Alternatively, one could check the condition  $p(E \cap F) = p(E) \cdot p(F)$ :  $1/4 = (1/2) \cdot (1/2)$ .

**Example 20.10.** The events E="more then one boy out of three kids" and F="the first baby is a girl" are not independent. Indeed,  $p(E \cap F) = p(\{GBB\}) = 1/8$ , whereas  $p(E) \cdot p(F) = (1/2) \cdot (1/2) = 1/4$ . So, p(E|F) = (1/8)/(1/2) = 1/4, i.e. E given F is twice less likely then E.

**Definition 20.11. Bernoulli trial** is an experiment with two outcomes: Success (probability p) and Failure (probability q = 1 - p). Examples: fair coin p = q = 1/2, biased coin p = 2/3, q = 1/3, etc.

**Theorem 20.12.** The probability of k successes in n independent Bernoulli trials with probability of success p is  $C(n,k) \cdot p^k \cdot q^{n-k}$ .

**Proof.** Each *n*-trail with k successes can be labelled by a string of k S's and (n-k) F's with the probability of such a string  $p^k q^{n-k}$ . There are C(n,k) such trials, the event E consists of all of them, therefore,  $p(E) = C(n,k)p^k q^{n-k}$ .

Homework assignments. (due Friday 03/16).

20A:Rosen4.5-6; 20B:Rosen4.5-10; 20C:Rosen4.5-16; 20D:Rosen4.5-26ac.