## **Discrete Probability and Counting**

A finite probability space is a set S and a function  $p: S \to R_{\geq 0}$  s.t.:

- $p(s) > 0 \forall s \in S$  and
- $\sum_{s \in S} p(s) = 1.$

We refer to S as the sample space, subsets of S as events and p as the probability distribution. The probability of an event  $A \subseteq S$  is  $\sum_{a \in A} p(a)$ .  $(p(\emptyset) = 0.)$ 

**Example**: Suppose we flip a fair coin. Saying the coin is fair implies that it is equally likely to flip H (heads) or T (tails) therefore p(H) = p(T) = 1/2.

If we assign all elements of S the same probability, as in the example above, then p is the *uniform distribution*.

**Example**: Suppose we flip a biased coin where the probability of H is twice as much as the probability of T. Since p(H) + p(T) = 1, this implies p(H) = 2/3 and p(T) = 1/3.

**Example:** Suppose we flip a fair coin twice. What is the probability of getting one H and one T? All the possible outcomes are  $\{HH, HT, TH, TT\}$ . Two out of the possible 4 outcomes give us one H and one T, each outcome has probability 1/4 therefore the total probability is 1/2

Suppose we flipped a fair coin n times. How many possible outcomes are there? There are two choices for each flip of the coin, so there are  $2^n$  possible outcomes. The probability of getting any one of these is  $1/2^n$ . (Where did we see this before?)

Now suppose we want to know the probability of getting exactly k Hs. We need to know how many of the  $2^n$  strings have exactly k Hs. In general, the number of ways to choose k things from n is given by:

 $\binom{n}{k} = n!/(n-k)!k!$ 

Where n! = n(n-1)(n-2)...1 and is read *n* factorial. We define 0! = 1. Note that  $\binom{n}{k} = \binom{n}{n-k}$ . These numbers are known as the *binomial coefficients*. Consider  $(x + y)^2 = x^2 + 2xy + y^2$ . The coefficients of this polynomial are  $\{1, 2, 1\}$  which are the numbers  $\binom{2}{0}$ ,  $\binom{2}{1}$ ,  $\binom{2}{2}$ . In general,  $(x + y)^n = \binom{n}{0}x^n + \binom{n}{1}x^{n-1}y + \binom{n}{2}x^{n-2}y^2 + \ldots + \binom{n}{n-1}xy^{n-1} + \binom{n}{n}y^n$ . **Example**: Suppose we flip a fair coin 10 times. What is the probability of getting exactly 4 *Hs*? First we compute  $\binom{10}{4} = 210$ . Then we compute the total number of outcomes  $2^{10} = 1024$ . Therefore the probability of getting exactly 4 *Hs* is  $210/1024 \approx .205$ 

Two events are *disjoint* if their intersection is empty.

**Example**: In the example of flipping 2 coins, the event A = 'getting exactly one H' and the event B = 'getting exactly 2Hs' are disjoint. But, A is not disjoint from the event C = 'getting exactly one T'. In fact, events A and C are the same in this case.

In general we have:  $p(A \cup B) + p(A \cap B) = p(A) + p(B)$ . Therefore, for disjoint events we have:  $p(A \cup B) = p(A) + p(B)$ . The first statement follows from the principle of *inclusion* - *exclusion* which states that  $|A \cup B| =$  $|A| + |B| - |A \cap B|$ .

**Example**: Say we flip a coin 10 times. What is the probability that the first flip is a T or the last flip is a T? The number of outcomes with the first flip T is  $2^9$ . The number of outcomes where the last flip is a T is  $2^9$ . The number of strings with both properties is  $2^8$ . Hence, the number of strings with either property is  $2^9 + 2^9 - 2^8 = 768$ .

Suppose we know that one event has happened and then want to ask about another. For two events A and B, the *conditional probability* of A relative to B is  $p(A|B) = p(A \cap B)/p(B)$  and read the probability of A given B.

**Example**: Suppose we flip a fair coin 3 times. Let *B* be the event that we have at least one *H* and *A* be the event of getting exactly 2 *H*s. What is the probability of *A* given *B*? In this case,  $(A \cap B) = A$ , p(A) = 3/8 (why?), p(B) = 7/8 (why?), and therefore p(A|B) = 3/7.

Notice that the definition of conditional probability also gives us the formula:  $p(A \cap B) = p(A|B)p(B)$ . For three events we have:  $p(A \cap B \cap C) = p(A|B \cap C)p(B|C)p(C)$ . (What is a general rule?)

We can also use conditional probabilities to find the probability of an event by breaking the sample space into disjoint pieces. If  $S = S_1 \cup S_2 \ldots \cup S_n$ 

and all pairs  $S_i$ ,  $S_j$  are disjoint then for any event A,  $p(A) = \sum p(A|S_i)p(S_i)$ .

**Example**: Suppose we flip a fair coin twice. Let  $S_1$  be the outcomes where the first flip is H and  $S_2$  be the outcomes where the first flip is T. What is the probability of A = getting 2 Hs? p(A) = (1/2)(1/2) + (0)(1/2) = 1/4.

Two events A and B are *independent* if  $p(A \cap B) = p(A)p(B)$ . This immediately gives: A and B are independent iff p(A|B) = p(A).

If  $p(A \cap B) > p(A)p(B)$  then A and B are positively correlated.

If  $p(A \cap B) < p(A)p(B)$  then A and B are negatively correlated.

**Example**: In the example of flipping 3 coins,  $p(A|B) \neq p(A)$  therefore these two events are not independent. Let *C* be the event that we get at least one *H* and at least one *T*. Let *D* be the event that we get at most one *H*. p(C) = 6/8, p(D) = 4/8, and  $p(C \cap D) = 3/8$  therefore events *C* and *D* are independent.

We say events  $A_1, \ldots A_n$  are mutually independent if for all subsets  $S \subseteq \{1, \ldots, n\}, p(\bigcap_{i \in S} A_i) = \prod p(A_i)$ . (What is an example of a set of mutually independent events?)