# 1. Probability and counting

### 1.2. Sample Spaces and Pebble World

The **sample space** S of an experiment = set of all possible outcomes of the experiment. Can be finite or infinite.

An **event** A is a subset of the S, and we say that A **occurred** if the actual outcome is in A.

not  $A = A^c$ 

Example:

 $A, B \subseteq S$ 

 $A \cup B$  if and only if at least one of A, B occurs.

 $A \cap B$  if and only if both of A, B.

De Morgan's laws:

 $(A \cup B)^c = A^c \cap B^c \mid A$  OR B not occur = A not occur AND B not occur  $(A \cap B)^c = A^c \cup B^c \mid A$  AND B not occur = A not occur OR B not occur

### 1.3 Naive definition of probability

$$P_{naive}(A) = rac{|A|}{|S|}$$

Requires S to be finite, with equal mass for each outcome.

### 1.4 How to count

Multiplication rule:

Consider experiment C = 2 sub-experiments A, B. A has a outcomes, B has b outcomes. Then C has ab outcomes.

Sampling without replacement:

n objects and making k choices from them, one at a time without replacement. Number of outcomes:  $n(n-1)\dots(n-k+1)$  for  $n\leq k$ .

A **permutation** of  $1, 2, \ldots, n$  is an arrangement of them in some order, e.g., 3,5,1,2,4 is a permutation of 1,2,3,4,5.

There are n! permutations of  $1, 2, \ldots, n$ 

**Binomial coefficient:**  $\binom{n}{k}$  = "n choose k" number of subsets of size k for a set of size n. For example,  $\binom{4}{2} = 6$ 

For  $k \leq n$ :

$$egin{pmatrix} n \ k \end{pmatrix} = rac{n(n-1)\dots(n-k+1)}{k!} = rac{n!}{(n-k)!k!}$$

For k > n:

$$\binom{n}{k} = 0$$

## 1.5 Story Proofs

Example: Vandermonde's identity

$$egin{pmatrix} n+m \ k \end{pmatrix} = \sum_{j=0}^k inom{m}{j} inom{n}{k-j}$$

Story proof:

Consider a group of m peacocks and n toucans, from which a set of size k birds will be chosen. There are  $\binom{n+m}{k}$  possibilities for this set of birds. If there are jpeacocks in the set, then there must be k-i toucans in the set. The right-hand side of Vandermonde's identity sums up the cases for *j*.

### 1.6 General definition of probability

A **probability space** = sample space S and probability function P. Event  $A \subseteq S$ . P(A) = [0, 1].

Axioms for probability function P:

- 1.  $P(\emptyset) = 0, P(1) = 1$
- 2. If  $A_1, A_2, \ldots$  are disjoint:

$$P(igcup_{j=1}^{\infty}A_j)=\sum_{j=1}^{\infty}P(A_j).$$

(" $A_i,A_j$  are disjoint" means that  $A_i\cap A_j=\emptyset$  if i
eq j)

The **frequentist view** of probability: if we say a coin has probability 1/2 of Heads, that means the coin would land Heads 50% of the time if we tossed it over and over and over (long-run frequency).

The **Bayesian** view of probability: represents a degree of belief about the event in question. We can assign probabilities to hypotheses even if it isn't possible to repeat the experiment over and over again.

#### **Properties of probability:**

 $\forall A, B,$ 

- 1.  $P(A^C) = 1 P(A)$
- 2. If  $A \subseteq B$ , then  $P(A) \le P(B)$ 3.  $P(A \cup B) = P(A) + P(B) P(A \cap B)$  inclusion-exclusion

Inclusion-exclusion for n events:

$$\forall A_1, \ldots A_n$$
,

$$P(igcup_{i=1}^n A_i) = \sum_i P(A_i) - \sum_{i < j} P(A_i \cap A_j) + \sum_{i < j < k} P(A_i \cap A_j \cap A_k) - \ldots + (-1)^{n+1} P(A_i \cap \ldots \cap A_n)$$

Can be simply used if  $A_i$  are symmetric else it is very tedious.

**NB:** Hypergeometric probability