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## 1.1 COUNTING

**Proposition (Basic principle of counting)** Suppose two independent experiments are performed, and there are  $m$  possible outcomes of the first experiment and  $n$  possible outcomes of the second experiment. Then the total possible outcomes of the two experiments combined is  $mn$ .

*Proof* Let  $(i, j)$  denote the case when the first experiment gives the  $i$ th outcome and the second experiment gives the  $j$ th outcome. Enumerating, we get

$$\begin{array}{cccc} (1, 1) & (1, 2) & \dots & (1, n) \\ (2, 1) & (2, 2) & \dots & (2, n) \\ \vdots & \vdots & \ddots & \vdots \\ (m, 1) & (m, 2) & \dots & (m, n) \end{array}$$

Since there are  $m$  rows and  $n$  columns, we have total  $mn$  entries. □

**Remark** This can be generalized to a finite number of experiments.

**Theorem (Binomial theorem)**

$$(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}.$$

*Proof (Inductive)* Homework. □

*Proof (Combinatorial)* Consider the product  $(x_1 + y_1)(x_2 + y_2) \cdots (x_n + y_n)$ .

First, note that the expansion consists of  $2^n$  terms, each being a product of  $n$  factors. Secondly, each product contains either  $x_j$  xor  $y_j$  for each  $j \in [n]$ .

For example,  $(x_1 + y_1)(x_2 + y_2) = x_1x_2 + x_1y_2 + y_1x_2 + y_1y_2$ .

Now, we can choose  $k$  of the  $x_j$ s and  $n - k$  of the  $y_j$ s in  $\binom{n}{k}$  ways, so there are precisely those many terms with  $k$   $x_j$ s and  $n - k$   $y_j$ s in the expansion.

Finally, letting  $x_j = x$  and  $y_j = y$  for each  $j \in [n]$ , we get the result. □

**Remark** This can be generalized to a finite number of experiments.



## 2.1 DISCRETE PROBABILITY SPACES

### Notations

Term	Description	Symbol/Idea	Coin toss Example
sample space	set of outcomes	$\Omega$	$\{H, T\}$
outcome	arbitrary outcome	$\omega \in \Omega$	$H$
event	subset of sample space	$E$	$\emptyset, \{H\}, \{T\}, \{H, T\}$
mutually exclusive events	events with empty intersection	$E_1 \cap E_2 = \emptyset$	$\{H\}$ and $\{T\}$
probability mass function	weightage of each outcome	$p : \Omega \rightarrow [0, 1]$ , with $\sum_{\omega} p(\omega) = 1$	$p(H) = \frac{1}{3}, p(T) = \frac{2}{3}$
probability	(of an event)	$\mathbb{P} : 2^{\Omega} \rightarrow [0, 1]$ , $\mathbb{P}(E) = \sum_{\omega \in E} p(\omega)$	$\mathbb{P}(\emptyset) = 0, \mathbb{P}(\{H, T\}) = 1$
random variable	a function	$X : \Omega \rightarrow \mathbb{R}$	$X(H) = 1, X(T) = 0$

**Definition (Probability axioms)** A non-negative valued function  $\mathbb{P}$  defined on the set of events is called a probability measure if the following hold.

1. (null empty set)  $\mathbb{P}(\emptyset) = 0$ .
2. (countable additivity) For any sequence of mutually exclusive events  $E_1, E_2, \dots$ , we have  $\mathbb{P}(\bigsqcup_{n=1}^{\infty} E_n) = \sum_{n=1}^{\infty} \mathbb{P}(E_n)$ .
3. (probability)  $\mathbb{P}(\Omega) = 1$ .

Draw Venn diagrams for all of the following.

**Proposition**  $\mathbb{P}(E^C) = 1 - \mathbb{P}(E)$ .

*Proof* Since  $E \cap E^C = \emptyset$ , by **Axiom 2** we have  $1 = \mathbb{P}(\Omega) = \mathbb{P}(E \sqcup E^C) = \mathbb{P}(E) + \mathbb{P}(E^C)$ .  $\square$

**Proposition** If  $E \subset F$ , then  $\mathbb{P}(E) \leq \mathbb{P}(F)$ .

*Proof* Note that  $F = E \sqcup (F \setminus E)$ . So by **Axiom 2** we have  $\mathbb{P}(F) = \mathbb{P}(E \sqcup (F \setminus E)) = \mathbb{P}(E) + \mathbb{P}(F \setminus E)$ . Therefore,  $\mathbb{P}(F) - \mathbb{P}(E) = \mathbb{P}(F \setminus E)$ , which is non-negative since probability is a non-negative set function.  $\square$

**Proposition (Inclusion-Exclusion)**  $\mathbb{P}(E \cup F) = \mathbb{P}(E) + \mathbb{P}(F) - \mathbb{P}(E \cap F)$ .

*Proof*

1.  $E \cup F = (E \setminus F) \sqcup (F \setminus E) \sqcup (E \cap F)$ , so  $\mathbb{P}(E \cup F) = \mathbb{P}(E \setminus F) + \mathbb{P}(F \setminus E) + \mathbb{P}(E \cap F)$ .
2.  $E = (E \setminus F) \sqcup (E \cap F)$ , so  $\mathbb{P}(E) = \mathbb{P}(E \setminus F) + \mathbb{P}(E \cap F)$ , and similarly
3.  $F = (F \setminus E) \sqcup (E \cap F)$ , so  $\mathbb{P}(F) = \mathbb{P}(F \setminus E) + \mathbb{P}(E \cap F)$ .

Combining the above,

$$\begin{aligned}\mathbb{P}(E \cup F) &= \mathbb{P}(E \setminus F) + \mathbb{P}(F \setminus E) + \mathbb{P}(E \cap F) \\ &= (\mathbb{P}(E) - \mathbb{P}(E \cap F)) + (\mathbb{P}(F) - \mathbb{P}(E \cap F)) + \mathbb{P}(E \cap F) \\ &= \mathbb{P}(E) + \mathbb{P}(F) - \mathbb{P}(E \cap F).\end{aligned}$$

□



## BIBLIOGRAPHY