

## Combinatorics

We want to choose  $k$  elements among  $n$  elements.

	Repetition not allowed	Repetition allowed
Order does not matter	$\frac{n!}{(n-k)!k!}$ <p>(<math>n</math> choices, then <math>n-1</math> choices, etc. and we stop at <math>n-k</math>, we also remove <math>n!</math> because order does not matter and <math>n!</math> is the number of permutations)</p>	$\binom{n-1+r}{n-1}$ <p>It's stars and bars method. We want <math>r</math> stars + <math>n</math> stars for each box. Then, we transform <math>n-1</math> of these stars into bars to separate the <math>r</math> stars into boxes.</p>
Order matters	$n!$ <p>(<math>n</math> choices, then <math>n-1</math> choices, etc.)</p>	$n^k$ <p>Cartesian product.</p>

## Probabilities

### Solving a probability problem

- list possible outcomes, define the probability space
- sometimes we keep a general  $\Omega$  and different  $\mathcal{F}$  depending on the point of view (colorblind/not colorblind, etc.)

### Terminology

- $\Omega$  is the **sample space**, containing all possible outcomes  $\omega$ .
- $\mathcal{F}$  is an **event space** (there are multiple event spaces!). It is a set of the subsets of  $\Omega$ . The powerset of  $\Omega$  includes all  $\mathcal{F}$ .  $|\mathcal{F}| = 2^{|\Omega|}$  only if  $\Omega$  is finite.

$\mathcal{F}$  is also called a sigma-algebra.

### Example for a fair die:

- **Sample space:**  $\{1, 2, 3, 4, 5, 6\}$
- **Events:**  $\{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{6\}, \{1, 2\}, \{1, 3\}, \dots, \{1, 2, 3, 4, 5, 6\}$  note that this is only for one throw!  $\{a, b\}$  is read "getting  $a$  or getting  $b$ ".
- **Event space:** all events

### Axioms

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Generalization of the OR between events:

$$P(A_1) + P(A_2) + P(A_3) - P(A_1 \cap A_2) - P(A_2 \cap A_3) - P(A_1 \cap A_3) + P(A_1 \cap A_2 \cap A_3)$$

$$\mathbb{P}(\cup_{i=1}^n A_i) = \sum_{r=1}^n (-1)^{r+1} \sum_{1 \leq i_1 < i_2 < \dots < i_r \leq n} (A_{i_1} \cap \dots \cap A_{i_r})$$

(les  $i_k$  doivent donc être différents)

Conditional probabilities:

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

### Law of total probability

Let  $\{B_i\}_{i=1}^{\infty}$  be pairwise disjoint events, and let  $A \subset \cup_{i=1}^{\infty} B_i$  then:

$$P(A) = \sum_{i=1}^{\infty} P(A \cap B_i) = \sum_{i=1}^{\infty} \mathbb{P}(A|B_i)\mathbb{P}(B_i)$$

### Bayes' Theorem:

$$\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}$$

but we can replace  $P(B)$  by what we know, law of total probability:

$$\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)} = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B|A)\mathbb{P}(A) + \mathbb{P}(B|A^c)\mathbb{P}(A^c)}$$

### Derangement

$$!n = \sum_{i=0}^n \frac{(-1)^i}{i!}, n \geq 0$$

Si on veut un élément non dérangé parmi  $n$ :

$$n \cdot \frac{1}{n} \cdot !(n-1)$$

Si on en veut  $r$ :

(comment on choisit les  $r$  bien rangés · le fait qu'ils soient bien rangés et que les restants soient dérangés)

$$\binom{n}{r} \cdot \frac{(n-r)!}{n!} \cdot !(n-r)$$

### Independence

If  $A$  and  $B$  are independent then  $\mathbb{P}(A|B) = \mathbb{P}(A)$ .

They are independent iff  $\mathbb{P}(A \cup B) = \mathbb{P}(A)\mathbb{P}(B)$ .

If two events are disjoint they can not be independent unless their respective probabilities are 0.

### Pairwise Independence

If you take any two events, they are independent.

### Random variable (notation)

We will use  $Y$  to denote a random variable.

We will use  $y$  to denote a specific value that  $Y$  can take.

$$\mathbb{P}(Y = y) \equiv p(y)$$

A probability distribution is a table or a graph that provides  $p(y) \forall y$ .

For everything to hold,  $\sum_y p(y) = 1$ .

### Expected Value

$$E(Y) = \sum_y yp(y)$$

### Binomial Distribution