## **Exercise 1.** Introduction

## **Binomial coefficients**

1. Let k,n be non negative integers. Give three definitions of  $\binom{n}{k}$ : an algebraic one, a combinatorial one, and its value.

Solution. The three definitions are

- (a) The coefficient in front of  $x^k$  in  $(1+x)^n$  (or the coefficient in front of  $a^k b^{n-k}$  in  $(a+b)^n$ ).
- (b) The number of ways to choose k elements in a set of n elements.
- (c) Is equal to

$$\frac{n!}{k! \left(n-k\right)!}$$

2. Prove that  $\begin{pmatrix} n \\ k \end{pmatrix} = \begin{pmatrix} n \\ n-k \end{pmatrix}$ .

Solution. We give three solutions

(a) Since  $(a+b)^n = (b+a)^n$ ,  $\binom{n}{k}$  is the coefficient in front of  $a^k b^{n-k}$  in  $(b+a)^n$  so it is the coefficient in front of  $b^{n-k}a^k$  in  $(b+a)^n$  so it is equal to

$$\left(\begin{array}{c}n\\n-k\end{array}\right).$$

(b) To choose k elements out of n is equivalent to discard n - k out of n.

(c) We have 
$$\binom{n}{n-k} = \frac{n!}{(n-k)!(n-(n-k))!} = \frac{n!}{k!(n-k)!} = \binom{n}{k}$$
.

3. Show that

$$\left(\begin{array}{c}n\\k\end{array}\right) + \left(\begin{array}{c}n\\k+1\end{array}\right) = \left(\begin{array}{c}n+1\\k+1\end{array}\right).$$

Solution. We give three solutions

- (a)  $\binom{n+1}{k+1}$  is the coefficient in front of  $x^{k+1}$  in  $(1+x)^{n+1}$ . Yet,  $(1+x)^{n+1} = (1+x)(1+x)^n = (1+x)(1+x)^n = (1+x)(1+x)^n$ = $(1+x)^n + x(1+x)^n$ . Thus  $\binom{n+1}{k+1}$  is the sum of the coefficient in front of  $x^{k+1}$  in  $(1+x)^n$  and the coefficient in front of  $x^k$  in  $(1+x)^n$ .
- (b) Let us consider the integers  $\{1, \dots, n+1\}$ . In order to choose k+1 elements in  $\{1, \dots, n+1\}$ , either one choose n+1 and then we need to choose k elements in  $\{1, \dots, n\}$  or one discards n+1 and then we need to choose k+1 elements in  $\{1, \dots, n\}$ .
- (c) We have

$$\binom{n}{k} + \binom{n}{k+1} = \frac{n!}{k!(n-k)!} + \frac{n!}{(k+1)!(n-k-1)!}$$
$$= \frac{n!}{(k+1)!(n-k)!}(k+1+n-k)$$
$$= \frac{(n+1)!}{(k+1)!(n-k)!} = \binom{n+1}{k+1}.$$

4. What is the value of  $\sum_{k=0}^{n} \binom{n}{k}$ ?

Solution. We give two solutions

- (a) We have  $\sum_{k=0}^{n} \binom{n}{k} = \sum_{k=0}^{n} \binom{n}{k} \mathbf{1}^{k} \mathbf{1}^{n-k} = (1+1)^{n} = 2^{n}.$
- (b)  $\sum_{k=0}^{n} \binom{n}{k}$  counts the number of subsets in a set with *n* elements. One has to choose if each element is included or not, thus there are 2 possibilities per element of the set:  $\sum_{k=0}^{n} \binom{n}{k} = 2^{n}$ .
- 5. Prove that

$$\sum_{\substack{k_1+k_2=k\\k_1,k_2\geq 0}} \binom{n_1}{k_1} \binom{n_2}{k_2} = \binom{n_1+n_2}{k}$$

Solution. We give two solutions

- (a)  $\sum_{\substack{k_1+k_2=k\\k_1,k_2\geq 0}} \binom{n_1}{k_1} \binom{n_2}{k_2}$  is the coefficient of  $x^k$  in  $(1+x)^{n_1} (1+x)^{n_2} = (1+x)^{n_1+n_2}$ , so it is equal to  $\binom{n_1+n_2}{k}$ .
- (b) Let us consider the integers  $\{1, \dots, n_1, \dots, n_1 + n_2\}$ . In order to choose k elements in  $\{1, \dots, n_1, \dots, n_1 + n_2\}$ , one needs to choose  $k_1$ , the number of elements to take from  $\{1, \dots, n_1\}$  and  $k_2$  the number of elements to take from  $\{n_1, \dots, n_1 + n_2\}$  (and of course  $k_1 + k_2 = k$ ) and then chose  $k_1$ , elements in  $\{1, \dots, n_1\}$  ( $\begin{pmatrix} n_1 \\ k_1 \end{pmatrix}$  possibilities) and  $k_2$  elements in  $\{n_1, \dots, n_1 + n_2\}$  ( $\begin{pmatrix} n_2 \\ k_2 \end{pmatrix}$  possibilities). This gives us the equality  $\sum_{\substack{k_1+k_2=k \\ k_1,k_2\geq 0}} \binom{n_1}{k_1} \binom{n_2}{k_2} = \binom{n_1+n_2}{k}$ .

## Stirling approximation

1. Recall the Stirling approximation.

**Solution.** Stirling's formula is  $n! = \sqrt{2\pi} n^{n+\frac{1}{2}} e^{-n} \left(1 + O\left(n^{-1}\right)\right)$ .

2. Show that

$$\frac{1}{2^{2n}} \left( \begin{array}{c} 2n \\ n \end{array} \right) \sim \frac{1}{\sqrt{\pi n}},$$

as  $n \to \infty$ .

Solution. This is a simple computation.

## Probabilities

1. Let  $A, B \subset (\Omega, \mathcal{A}, \mathbb{P})$ , be two events. What does it means that they are independent?

**Solution.** It means that  $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ .

2. What is the definition of the conditional probability  $\mathbb{P}(A|B)$ ? What is the value of  $\mathbb{P}(A|B)$  if A and B are independent ?

**Solution.** We have  $\mathbb{P}(A|B) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}$  which is equal to  $\mathbb{P}(A)$  if A and B are independent.

3. Let X be a non negative random variable. State and prove the Markov inequality.

**Solution.** The Markov inequality is the fact that for any  $a \ge 0$ ,

$$\mathbb{P}\left(X \ge a\right) \le \frac{\mathbb{E}\left(X\right)}{a}$$

The proof goes as follows:  $a \mathbb{1}_{X \ge a} \le X$  since X is non negative and computing the expectation, one gets the inequality.

4. Give the definition of a (discrete time) Markov process.

**Solution.** A random process  $(X_n)_{n \in \mathbb{N}}$  such that for any  $n \in \mathbb{N}$ , any  $m \in \mathbb{N}$ , any  $(x_1, \dots, x_{n+m})$ ,

$$\mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid (X_1, \cdots, X_n) = (x_1, \cdots, x_n)\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right) = \mathbb{P}\left((X_{n+1} = x_{n+1}, \cdots, X_{n+m} = x_{n+m}) \mid X_n = x_{n+1}, \cdots, x_n\right)$$

5. Let G be a general graph, explain what a simple random walk on G is.

**Solution.** A Markov process which jumps at each time, independently from the past, uniformly to one of its neighbours.

Recall that a simple random walk on a graph is called *recurrent* if it returns to the starting point with probability 1, and *transient* otherwise. Recall that a simple random walk  $(S_n)_{n\geq 0}$  on a connected graph G, starting from  $v \in G$ , is recurrent if and only if

$$\sum_{n=0}^{\infty} \mathbb{P}\left(S_n = v\right) = \infty. \tag{0.1}$$

*Remark.* In the course you saw that a simple random walk  $(S_n)_{n\geq\infty}$  is recurrent if and only if  $\mathbb{E}[N_d] = \infty$  where  $N_d$  is the number of visits at the starting point v. The relation with the statement above is obtained using the relation  $N_d = \sum_{n=0}^{\infty} \mathbf{1}_{\{S_n = v\}}$ , and using the linearity of the expectation:

$$\mathbb{E}[N_d] = \sum_{n=0}^{\infty} \mathbb{E}[\mathbf{1}_{\{S_n=v\}}] = \sum_{n=0}^{\infty} \mathbb{P}\left(S_n=v\right).$$

**Exercise 2.** Recurrence/transience theorem for simple random walks on the square lattice  $\mathbb{Z}^d$ ,  $d \ge 1$ . Let  $\left(S_n^{(d)}\right)_{n\geq 0}$  be the simple random walk on  $\mathbb{Z}^d$  such that  $S_0^{(d)} = 0$ .

1. d = 1 Use Stirling's formula<sup>1</sup> to show that, in one dimension,

$$\mathbb{P}\left(S_{2n}^{(1)}=0\right)\sim\frac{1}{\sqrt{\pi n}}.$$

Deduce that  $(S_n^{(1)})_{n>0}$  is recurrent.

**Solution.** In order for the simple random walk on  $\mathbb{Z}$  to come back to 0 in 2n steps, it must make n positive steps and n negative steps. Thus among the  $2^n$  possible walks (at each step, the walk has 2 choices), the number of walks coming back to the origin in 2n steps is equal to the number of ways to choose n positive steps in the 2n total steps. So

$$\mathbb{P}\left(S_{2n}^{(1)} = 0\right) = \frac{1}{2^{2n}} \begin{pmatrix} 2n \\ n \end{pmatrix} = \frac{(2n)!}{(n!)^2 2^{2n}}$$

Using Stirling's formula on the factorials immediately gives the result. The recurrence property follows from (0.1).

**2.** d = 2 The goal is to prove that the simple random walk on  $\mathbb{Z}^2$  is recurrent.

1. By enumerating the different cases, show that

$$\mathbb{P}\left(S_{2n}^{(2)}=0\right) = \left(\frac{1}{2^{2n}} \left(\begin{array}{c}2n\\n\end{array}\right)\right)^2.$$
(0.2)

**Solution.** Among the  $4^{2n}$  possible walks (4 choices at each step), the number of walks coming back to the origin in 2n steps can be obtained by

$$\begin{cases} \text{choosing } 2j \text{ steps in the } x \text{ direction} & \begin{pmatrix} 2n \\ 2j \end{pmatrix} \\ \text{choosing } j \text{ positive steps among these } 2j \text{ steps} & \begin{pmatrix} 2j \\ 2j \\ j \end{pmatrix} \\ \text{choosing } n-j \text{ positive steps among the } 2(n-j) & \begin{pmatrix} 2(n-j) \\ n-j \end{pmatrix} \\ \text{remaining steps (in } y \text{ direction)} \end{cases}$$

<sup>1</sup>Stirling's formula is  $n! = \sqrt{2\pi} n^{n+\frac{1}{2}} e^{-n} \left( 1 + O(n^{-1}) \right)$ .

Thus, the number of such walks is equal to

$$\begin{split} \sum_{j=0}^{n} \binom{2n}{2j} \binom{2j}{j} \binom{2(n-j)}{n-j} &= \sum_{j=0}^{n} \frac{(2n)!}{(2j)! (2n-2j)!} \frac{(2j)!}{j!j!} \frac{(2(n-j))!}{(n-j)! (n-j)!} \\ &= \frac{(2n)!}{n!n!} \sum_{j=0}^{n} \frac{n!n!}{j!j! (n-j)! (n-j)!} \\ &= \frac{(2n)!}{n!n!} \sum_{j=0}^{n} \binom{n}{j} \binom{n}{n-j} \end{split}$$

The last sum is equal to  $\frac{(2n)!}{n!n!}$ . Indeed either one can see that this is the coefficient of  $x^n$  in  $(1+x)^n (1+x)^n$  which is equal also to the coefficient of  $x^n$  in  $(1+x)^{2n}$ , or one can use the following combinatorial proof: in order to pick n elements in a bag of 2n elements, I split the bag in 2 smaller bags of equal size n (in an arbitrary way) and then I pick j elements in the first bag and n-j in the second bag.

Thus  $\mathbb{P}\left(S_{2n}^{(2)}=0\right) = \left(\frac{1}{2^{2n}}\frac{(2n)!}{n!n!}\right)^2$ , which is what we needed to prove.

1. Observe that  $\mathbb{P}\left(S_{2n}^{(2)}=0\right)$  is equal to  $\mathbb{P}\left(S_{2n}^{(1)}=0\right)^2$ . Find a probabilistic proof of Equation (0.2). **Solution.** The intuition behind the equality

$$\mathbb{P}\left(S_{2n}^{(2)}=0\right)=\mathbb{P}\left(S_{2n}^{(1)}=0\right)^2$$

is that one can represent  $(S_n^{(2)})_{n\geq 0}$  using two independent uni-dimensional random walks. Beginning at the origin, suppose at every step we do SRW in the x and y directions independently. Then we will move diagonally in  $\mathbb{Z}^2$ , and the resulting law of the walk in the rotated diagonal lattice is precisely that of a 2 dimensional simple random walk. Then we return to the origin in 2n steps if and only if the independent 1 dimensional SRWs both come back to zero, so we get the square of the one dimensional estimate.

Equivalently, one could have considered the projection of  $S_k = (X_k, Y_k)$  on the x and y axis. The projection  $X_k$  is clearly not a simple random walk since it stays sometimes at the same place. Yet  $X_k + Y_k$  and  $X_k - Y_k$  are two processes which always either increase or decrease by 1. Besides  $\{S_k = 0\} = \{X_k + Y_k = 0 \text{ and } X_k - Y_k = 0\}$ . At last, it is easy to see that  $X_k + Y_k$  and  $X_k - Y_k$  are two independent random walks (consider the way  $(X_k + Y_k, X_k - Y_k)$  moves from time k to time k + 1).

2. Deduce from Equation (0.2) that  $(S_n^{(2)})_{n>0}$  is recurrent.

**Solution.** From the part 1. we deduce that  $\mathbb{P}\left(S_{2n}^{(2)}=0\right) \sim \frac{1}{\pi n}$ . The recurrence property follows from (0.2).

**3.** d = 3 By a simple enumeration argument, show that

$$\mathbb{P}\left(S_{2n}^{(3)}=0\right) = \frac{1}{2^{2n}} \left(\begin{array}{c} 2n\\ n \end{array}\right) \sum_{j,k\geq 0\atop j+k\leq n} \left(\frac{n!}{3^n k! j! (n-k-j)!}\right)^2$$

and deduce that a simple random walk on  $\mathbb{Z}^3$  is transient.

**Solution.** There exists  $\frac{1}{6^{2n}}$  different paths that the random walk can follow during the first 2n steps (it has 3 choices at each step). We need to compute the number of paths of length 2n in  $\mathbb{Z}^3$  which begin and come back to 0. We need to choose :

 $\begin{cases} 2k \text{ times (among the } 2n) \text{ at which the path will go in the } x \text{ direction} & \begin{pmatrix} 2n \\ 2k \end{pmatrix} \\ 2j \text{ times (among the } 2n - 2k \text{ left) at which the path will go in the } y \text{ direction} & \begin{pmatrix} 2n \\ 2k \end{pmatrix} \\ 2n - 2k \\ 2j \end{pmatrix} \\ k \text{ times (among the } 2k) \text{ at which the path will go "up" in the } x \text{ direction} & \begin{pmatrix} 2n \\ 2k \end{pmatrix} \\ k \end{pmatrix} \\ j \text{ times (among the } 2j) \text{ at which the path will go "up" in the } y \text{ direction} & \begin{pmatrix} 2j \\ 2j \end{pmatrix} \\ j \end{pmatrix} \\ n - k - j \text{ times (among the } 2(n - k - j)) \text{ at which the path will go "up"} & \begin{pmatrix} 2n \\ 2k \end{pmatrix} \\ j \end{pmatrix} \\ n - k - j \text{ times (among the } 2(n - k - j)) \text{ at which the path will go "up"} \end{cases}$ 

This gives a number of paths equal to:

$$\sum_{j,k\geq 0\atop j/k\leq n} \binom{2n}{2k} \binom{2n-2k}{2j} \binom{2k}{k} \binom{2j}{j} \binom{2n-2k-2j}{n-k-j},$$

which after a little massage gives

$$\binom{2n}{n} \sum_{\substack{j,k \ge 0\\j/k \le n}} \left(\frac{n!}{k!j!(n-k-j)!}\right)^2,$$

and thus

$$\mathbb{P}\left(S_{2n}^{(3)}=0\right) = \frac{1}{6^{2n}} \left(\begin{array}{c}2n\\n\end{array}\right) \sum_{j,k \ge 0 \atop j/k \le n} \left(\frac{n!}{k!j!(n-k-j)!}\right)^2 = \frac{1}{2^{2n}} \left(\begin{array}{c}2n\\n\end{array}\right) \sum_{j,k \ge 0 \atop j/k \le n} \left(\frac{n!}{3^nk!j!(n-k-j)!}\right)^2.$$

For the assertion about the transience of the random walk, we need to show that  $\sum_{n} \mathbb{P}\left(S_{2n}^{(3)}=0\right) < \infty$ : we need to give an upper bound on  $\mathbb{P}\left(S_{2n}^{(3)}=0\right)$  which is summable. The first part  $\frac{1}{2^{2n}}\begin{pmatrix}2n\\n\end{pmatrix}$  was already studied it is  $O\left(\frac{1}{\sqrt{n}}\right)$ . It remains to bound  $\sum_{\substack{j,k\geq 0\\j/k\leq n}} \left(\frac{n!}{3^nk!j!(n-k-j)!}\right)^2$  and in particular  $\frac{n!}{k!j!(n-k-j)!}$ . Let us remark that if a < b then  $a!b! \ge (a+1)!(b-1)!$  since it is equivalent to  $b \ge a+1$ . Thus a!b!c! decreases when the distance between any two of a; b; c decreases. We conclude that  $\frac{n!}{k!j!(n-k-j)!}$  is maximized among the cases where j,k, n-j-k are of order n/3. Thus  $\frac{n!}{3^nk!j!(n-k-j)!} \le O\left(3^{-n}\frac{n!}{(\lfloor n/3 \rfloor!)^3}\right) = O\left(n^{-1}\right)$  using Stirling formula.

$$\mathbb{P}\left(S_{2n}^{(3)} = 0\right) \leq \frac{c}{n^{3/2}} \sum_{j,k \geq 0 \atop j+k \leq n} \frac{n!}{3^n k! j! \left(n-k-j\right)!} = \frac{c}{n^{3/2}}$$

since the sum of the multinomial coefficients is precisely  $3^n$ . This allows us to conclude about the transience of the random walk in dimension 3.

*Remark.* Let us remark that the brutal majoration which would consist in majoring  $\left(\frac{n!}{3^n k! j! (n-k-j)!}\right)^2$  by  $O\left(n^{-2}\right)$  and the sum by the number of elements (of order  $(O(n^2))$ ) times  $O\left(n^{-2}\right)$  would have given us a majoration  $\mathbb{P}\left(S_{2n}^{(3)}=0\right) \leq \frac{c}{n^{1/2}}$  and would have not helped us.

**4.**  $d \ge 3$  Prove that it follows from the previous results that  $\mathbb{Z}^d$  is transient for d > 3.

**Solution.** Given an SRW on  $\mathbb{Z}^d$  for d > 3, consider its projection  $S_n$  to the first three coordinates. This has a law of a Markov random walk on  $\mathbb{Z}^3$  started at the origin which at every step can move to one of its 6 neighbours with probability  $\frac{1}{2d}$ , or stay at the same point with probability  $1 - \frac{6}{2d}$ . But one obtains a SRW in  $\mathbb{Z}^3$  by disregarding the steps where the first three coordinates do not move. So our SRW on  $\mathbb{Z}^d$  does not return to zero in its first three coordinates infinitely often, let alone to the origin.