



## 1. Discrete-time random walk

A drunkard moves along a line by making each second a step to the right or to the left with equal probability. Thus his possible positions are the integers  $-\infty < n < \infty$ . We assume that initially  $n = 0$  and want to determine the probability  $p_n(r)$  after  $r$  steps.

### 1.1. Combinatorial derivation

Derive the probability  $p_n(r)$  by combinatorial arguments.

Hint: How many different path lead to the position  $n$  after  $r$  steps?

### 1.2. Solve by addition of variables

Each step  $j = 1, 2, \dots, r$  corresponds to a stochastic variable  $\hat{x}_j$  taking on the values 1 and  $-1$  with equal probability  $1/2$ . The position after  $r$  steps is then given by

$$\hat{n}_r = \hat{x}_1 + \hat{x}_2 + \dots + \hat{x}_r.$$

The steps and thus the variables  $\hat{x}_j$  are mutually independent.

#### 1.2.1. Average

Derive the average  $\langle \hat{n}_r \rangle$  and the variance  $\langle \hat{n}_r^2 \rangle$ . Discuss the  $r$ -dependence of these variables and compare to that of a deterministic (sober) movement.

#### 1.2.2. Characteristic Function

Calculate the characteristic function  $G_{\hat{n}_r}(k)$ . Determine the probability  $p_n(r)$  from this expression. What is the advantage of this method compared to the combinatorial derivation?

Hint:

$$(a + b)^n = \sum_{k=0}^n \binom{n}{k} a^{n-k} b^k$$

### 1.3. Asymmetric random walk

Now assume that a step to the left has the probability  $q$  and to the right  $1 - q$ . Find  $p_n(r)$  for this case.

## 2. Compound distribution

Let  $\hat{x}_j$  be an infinite set of independent stochastic variables with identical distributions  $P_{\hat{x}}(x)$  and characteristic function  $G_{\hat{x}}(k) = \int_{-\infty}^{\infty} P_{\hat{x}}(x) e^{ikx} dx$ . Let  $\hat{r}$  be a random non-negative integer with distribution  $p_r$  and probability generating function  $f(z) = \langle z^{\hat{r}} \rangle = \sum_{r=0}^{\infty} p_r z^r$ . Then the sum

$$\hat{y} = \hat{x}_1 + \hat{x}_2 + \dots + \hat{x}_{\hat{r}}$$

is a random variable, where we set  $\hat{y} = 0$  for  $\hat{r} = 0$ . Show that its characteristic function  $G_{\hat{y}}(k)$  fulfills

$$G_{\hat{y}}(k) = f(G_{\hat{x}}(k)).$$

The distribution of  $\hat{y}$  is the so-called *compound distribution*.

### 2.1. Example

Assume that  $\hat{x}_i$  are Bernoulli variables, i.e.  $P(\hat{x}_i = 1) = p = 1 - q$  and  $P(\hat{x}_i = -1) = q$  like in the asymmetric random walk in 1.3, and  $\hat{r}$  obeys Poisson statistics. How does the compound distribution  $P_{\hat{y}}$  then look like?

## 3. Branching Process

We now consider a so-called *branching process* representing a kind of chain reaction occurring in successive generations. Initially, for the zeroth generation, we start with a single event.

This event causes in the first generation  $\hat{r}$  further events, where  $\hat{r}$  is a random non-negative integer with distribution  $p_r$ . If  $\hat{r} = 0$ , the process stops. Otherwise, every direct descendant causes in the next (second) generation further events according to the same distribution  $p_r$  and so on. The events of each generation act independently of each other.

Whereas an obvious example for such a process are nuclear chain reactions, originally this process has been invented by F. Galton to study the survival of family names (hence the name *Galton-Watson process*). Another example is the cascading failure of components like power lines which is induced by an overloading occurring after the initial failure of a line.

### 3.1. Recursion relation

Let  $\hat{y}_n$  be the number of events, i.e., the size of the  $n$ -th generation with generating function  $f_n(z)$ . In particular,  $y_0 = 1$  in the 0th generation. Furthermore,  $f_1(z) = f(z)$ , where  $f(z) = \langle z^{\hat{r}} \rangle$  is the generating function corresponding to the variable  $\hat{r}$ . Derive from the description of the branching process, the recursive relation

$$f_n(z) = f(f_{n-1}(z)). \quad (1)$$

### 3.2. Probability for a finite cascade

An important question is whether the branching process continues forever or stops after a finite number of generations, i.e., whether  $\hat{y}_n = 0$  for an arbitrary  $n$ .

#### 3.2.1. Recursion formula

How can the probability  $q_n := P(\hat{y}_n = 0)$  that the process stops at or before the  $n$ -th generation be calculated from the generating function  $f_n(z)$ ? Derive from Eq. (1) a recursion formula for the  $q_n$ . What is the initial condition  $q_1$ ?

#### 3.2.2. Termination of branching process

Excluding the extreme cases  $p_0 = 0$  and  $p_0 = 1$  derive from the form of the generating function  $f(z)$  that the  $q_n$  approach a limiting value  $q \leq 1$  satisfying

$$q = f(q). \quad (2)$$

Show geometrically that equation (2) has the unique solution  $q = 1$  if and only if the first derivative  $f'(1)$  of the characteristic function satisfies  $f'(1) \leq 1$ .

Hint: The characteristic function is convex. Why?

### 3.2.3. Mean size of generations

Show that the expected size of the  $n$ -th generation is given by

$$\langle \hat{y}_n \rangle = f'_n(1).$$

Derive a chain rule for  $\langle \hat{y}_n \rangle$  and show that

$$\langle \hat{y}_n \rangle = \mu^n,$$

where  $\mu$  is the mean value of  $\hat{r}$ . Why is it intuitively clear that the branching process dies out for  $f'(1) < 1$ ?