

# 1 Extreme value statistics

- Random distribution  $p(x)$  in  $[0, \infty[$
- Largest variable  $x_{\max}$  out of  $N$  random variables

$$\int_{x_{\max}}^{\infty} p(x) dx \sim \frac{1}{N} \quad \rightsquigarrow \quad \text{estimate for } x_{\max}$$

More precisely, we get:

$$M_N(x) = N [1 - P(x)]^{N-1} p(x); \quad P(x) = \int_x^{\infty} p(y) dy$$

a)  $p(x) = e^{-x}$

$$\int_{x_{\max}}^{\infty} e^{-x} dx = e^{-x_{\max}} \approx \frac{1}{N}$$

$$\rightsquigarrow x_{\max} \approx \ln(N)$$

b)  $p(x) = \mu x^{-(1+\mu)} \quad x > 1; \mu > 0$

$$\int_{x_{\max}}^{\infty} \mu x^{-(1+\mu)} dx = x_{\max}^{-\mu} \sim \frac{1}{N}$$

$$\Rightarrow x_{\max} \sim N^{\frac{1}{\mu}}$$

c)  $p(x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{else} \end{cases}$

$$\int_{x_{\max}}^1 dx = (1 - x_{\max}) \sim \frac{1}{N}$$

$$\rightsquigarrow x_{\max} \sim 1 - \frac{1}{N}$$

## Perfectly absorbing boundary conditions

Diffusion equation

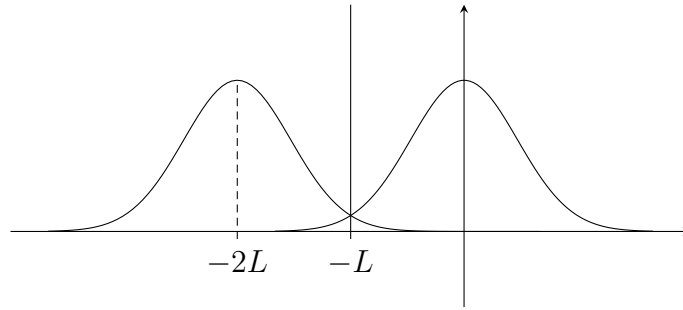
$$\partial_t p(x, t) = D \partial_x^2 p(x, t)$$

We consider the initial condition

$$p(x, 0) = M \delta(x)$$

and boundary conditions:

$$p(-L, t) = 0$$



blabla

For the case of two boundary conditions, we have to add an imaginary source at  $x = 2L$ . After a while, however the imaginary random walk may cross the boundary at  $\pm L$ . Therefore we need multiple imaginary walks.

In particular, we need negative image at  $x = \pm 2L$ , positive at  $x \pm 4L$  and so forth.

Therefore:  $p(x, t) = \frac{M}{\sqrt{4\pi Dt}} \sum_{n=-\infty}^{\infty} \left[ -\exp\left(-\frac{(x+(4n-2)L)^2}{4Dt}\right) + \exp\left(-\frac{(x+4nL)^2}{4Dt}\right) \right]$

where  $-L < x < L$ . Nummer 3:  $p(l) = \rho e^{-\rho l}$   $\rho L = l$  has to be taken into account