Prof. Dr. Joachim Krug Institut für Theoretische Physik Wintersemester 2011/12

# Probability theory and stochastic processes for physicists

# **Problem 4: Stable laws and renormalization**<sup>1</sup>

Consider two independent real random variables  $X_1$  and  $X_2$  drawn from the same pdf f(x), which is assumed for simplicity to have zero mean. The pdf  $f_{\Sigma}(y)$  of the sum  $Y = X_1 + X_2$ is then given by

$$f_{\Sigma}(y) = \int_{-\infty}^{\infty} dx \ f(y-x)f(x).$$
(1)

A pdf f is called a *stable law* if  $f_{\Sigma}$  is identical to f under appropriate rescaling of the argument, that is if

$$\mathcal{R}_b[f](y) \equiv bf_{\Sigma}(by) = b \int_{-\infty}^{\infty} dx \ f(by - x)f(x) = f(y)$$
<sup>(2)</sup>

for a suitably chosen scale factor b. Equation (2) can be viewed as a fixed point condition for a simple renormalization transformation  $\mathcal{R}_b$  acting on the space of probability distributions. Rewrite the fixed point condition (2) using the generating function

$$G_X(k) = \int_{-\infty}^{\infty} dx \ e^{ikx} f(x).$$
(3)

Show that the Gaussian and the Cauchy distributions

$$f_G(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}, \quad f_C(x) = \frac{1}{x^2 + \pi^2}$$
 (4)

are fixed points of (2), and identify the corresponding values of b. Show more generally that (2) is solved by the one-parameter family of (symmetric) Lévy stable laws with generating function  $e^{-C|k|^{\alpha}}$ , where  $0 < \alpha \leq 2$ .

## Problem 5: Zipf's law for random texts

Zipf's law states that the number N(x) of distinct words that occur with frequency x in a text is proportional to  $x^{-\alpha}$ , where  $\alpha \approx 2$ . Here we consider random texts, which are random uncorrelated sequences consisting of m different letters and one space sign which separates different words. All letters occur with equal probability q, and the space sign with probability  $q_s = 1 - mq$ . Clearly in this model all words of the same length l occur with the same probability. Show that both the probability of occurrence of a given word, and the number of distinct words of length l depend exponentially on l, and deduce from this the power law  $N(x) \sim x^{-\alpha}$ . Investigate the behavior of  $\alpha$  for large m and small  $q_s$ .

<sup>&</sup>lt;sup>1</sup>See Excercise 12.1. in *Statistical Mechanics: Entropy, Order Parameters, and Complexity* by J.P. Sethna, and Sect. 2.3 of *Critical Phenomena in the Natural Sciences* by D. Sornette.

#### Problem 6: Fitting power law distributions

In many applications one is faced with the task of deciding whether a sequence of N data points  $X_1, ..., X_N$  is consistent with an underlying power law probability distribution, and to determine the exponent. We assume here that the  $X_i$  are real numbers with  $X_i \ge 1$ , and the hypothetical probability density is of *Pareto* type,

$$f(x) = \alpha x^{-(\alpha+1)} \tag{5}$$

with the corresponding distribution function

$$F(x) = \int_{1}^{x} dx' f(x') = 1 - x^{-\alpha}.$$
 (6)

a.) By rank ordering the data points are ordered by increasing size, such that the maximal value  $X_{\text{max}}$  has rank  $R(X_{\text{max}}) = N$  and the minimal value has rank  $R(X_{\min}) = 1$ . Show that quite generally, for independent random variables and large N,

$$\frac{R(X_i)}{N} \to F(X_i).$$

b.) To determine the exponent  $\alpha$  from the data, one often fits a straight line to a doublelogarithmic plot of a histogram representing f(x) or 1 - F(x), or to the rank  $R(X_i)$ . A more accurate method is based on an estimate of the *likelihood* of the value of  $\alpha$ , given the data  $X_1, ..., X_N$ . A simple ansatz for the likelihood is<sup>2</sup>

$$\mathcal{L}(\alpha|X_1,...,X_N) = \ln\left[\prod_{i=1}^N f(X_i)\right].$$
(7)

Derive a formula for the most likely value of  $\alpha$  by maximizing (7) under the hypothesis that the distribution is given by (5).

c.) Generate numerically a sample of N = 10000 power-law distributed random variables by transforming uniform random numbers according to (6). Then try to extract the exponent from the data using (i) a linear fit to  $\ln R(X_i)$  versus  $\ln X_i$ , and (ii) the maximum likelihood formula from part b.). Compare the results.

### Problem 7: Combining power law random variables

Let  $X_1, X_2 \ge 1$  denote i.i.d. random variables with Pareto distribution (6).

- a.) Compute the density  $f_Y(y)$  of the sum  $Y = X_1 + X_2$ , and show that it behaves as  $f(y) \approx 2\alpha Y^{-(\alpha+1)}$  for large y.
- b.) Compute the density  $f_Z(z)$  of the product  $Z = X_1 X_2$ , and show that it behaves as  $f(z) \sim \ln(z) z^{-(\alpha+1)}$  for large z.

<sup>&</sup>lt;sup>2</sup>M.E.J. Newman, Cont. Phys. **46**, 323 (2005); A. Clauset, C.R. Shalizi and M.E.J. Newman, SIAM Review **51**, 661 (2009).