

# Extreme value theory midterm exam, 15th Apr 2024

1. (a) What does it mean that a function is slowly varying at infinity?  
 (b) Which of the functions

$$f(x) = (\log x)^2, \quad g(x) = \frac{x^2}{\log x}, \quad h(x) = \frac{\log x}{x^2}$$

defined for  $x > 0$  are slowly varying at infinity?

2. Let  $\tau = \inf\{t > 0 : B_t = 1\}$  be the hitting time of level 1 by the standard Brownian motion  $B_t$  which starts from  $B_0 = 0$ . Show that  $\tau$  has the Lévy distribution, that is, its density is  $(2\pi y^3)^{-1/2} \exp(-1/(2y))$  for  $y > 0$ .

*Hint:* It can be used without proof that as a consequence of the reflection principle

$$\mathbf{P}(\max\{B_s : s \in [0, t]\} > x) = 2 \left(1 - \Phi\left(\frac{x}{\sqrt{t}}\right)\right)$$

where  $\Phi(x) = \int_{-\infty}^x (2\pi)^{-1/2} \exp(-y^2/2) dy$  is the standard normal distribution function. The distribution function of  $\tau$  can be directly expressed by the probability in the equality above.

3. Let  $X_1, X_2, \dots$  be an iid. sequence of negative binomial random variables with parameters  $p \in (0, 1)$  and  $m = 2$ , that is,

$$\mathbf{P}(X = k) = (k + 1)p^2(1 - p)^k$$

for  $k = 0, 1, 2, \dots$  and let  $M_n = \max(X_1, \dots, X_n)$ . Show that there is no such normalization under which the sequence of maxima  $M_n$  has a non-degenerate limit law.

*Hint:* Show that the necessary condition

$$\lim_{x \uparrow x_F} \frac{\overline{F}(x+)}{\overline{F}(x)} = 1$$

fails to hold along a sequence of integers where  $\overline{F}(x) = 1 - F(x)$  is the tail probability function of  $X_i$ . More precisely  $\overline{F}(k+1)/\overline{F}(k)$  converges to  $1 - p$  for the integers  $k \rightarrow \infty$ .

4. Let  $U_1, U_2, \dots$  be a sequence of independent and identically distributed uniform random variables on  $[0, 1]$ . Define  $X_n = e^{U_n}$  for all  $n = 1, 2, \dots$  and let  $M_n = \max(X_1, \dots, X_n)$ . Which non-trivial limit distribution does the renormalized sequence of  $M_n$  converge to? Under what normalization?
5. Let  $X_1, X_2, \dots$  be random variables defined on the probability space  $(\Omega, \mathcal{A}, \mathbf{P})$  which have the same distribution function  $F(x)$  but we do not assume anything about their dependence structure. Let  $M_n := \max_{1 \leq i \leq n} |X_i|$ . Suppose that there is an  $s > 0$  such that  $\int_{-\infty}^{\infty} e^{s|x|} dF(x) < \infty$ , that is,  $\mathbf{E}(e^{s|X_i|}) < \infty$ . Prove that for any sequence  $b_n$  that goes to  $\infty$  and for fixed  $\delta > 0$ ,

$$\lim_{n \rightarrow \infty} \mathbf{P}((b_n \log n)^{-1} |M_n| > \delta) = 0.$$

*Hint:* Use the following inequality

$$\mathbf{P}\left(\max_{1 \leq i \leq n} |X_i| > \lambda\right) = \mathbf{P}(\cup_{i=1}^n \{|X_i| > \lambda\}) \leq \sum_{i=1}^n \mathbf{P}(|X_i| > \lambda) = n\mathbf{P}(|X_i| > \lambda)$$

and a Markov type inequality.