

Problem set 1, Advanced risk theory, 24.1.2020

Exercise 1.1. Consider functions $f: \mathbb{R} \rightarrow (0, \infty)$ and $g: \mathbb{R} \rightarrow (0, \infty)$. The functions are said to be *asymptotically equivalent* (at infinity) if

$$(1) \quad \lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = 1.$$

Relation (1) is often abbreviated using notation $f(x) \sim g(x)$, as $x \rightarrow \infty$.

- a) Suppose $g(x) = e^{-Rx}$ for a fixed $R > 0$ and all $x \in \mathbb{R}$. Show that if $f(x) \sim g(x)$, as $x \rightarrow \infty$, then $-\log f(x) \sim Rx$, as $x \rightarrow \infty$.
- b) Let f and g be as in Part a). Give an example of f such that $-\log f(x) \sim Rx$ holds, but $f(x) \sim g(x)$ does not hold, as $x \rightarrow \infty$.

Exercise 1.2. The *exponential index* $\mathcal{E}(\xi)$ of random variable ξ is defined in risk theory as

$$\mathcal{E}(\xi) = \sup \left\{ s \geq 0 : \mathbb{E} \left(e^{s\xi} \right) < \infty \right\} = \liminf_{z \rightarrow \infty} \frac{-\log \mathbb{P}(\xi > z)}{z}.$$

Let $a \in \mathbb{R}$. Let c_ξ be the cumulant generating function of ξ so that $\mathcal{E}(\xi) = \sup \{s \geq 0 : c_\xi(s) < \infty\}$.

Give an example of a distribution of ξ for which $\mathcal{E}(\xi) \in (0, \infty)$ and $c_\xi(\mathcal{E}(\xi)) = a$. Hint: You can make a slight modification to a distribution with density $f(x) = Ce^{-x}/(1+x^3)$ for $x > 0$ to match the value of $c_\xi(\mathcal{E}(\xi))$ with a .

Exercise 1.3. The *moment index* $\mathbb{I}(\xi)$ of random variable ξ is defined in risk theory as

$$\mathbb{I}(\xi) = \sup \left\{ s \geq 0 : \mathbb{E} \left((\xi^+)^s \right) < \infty \right\} = \liminf_{z \rightarrow \infty} \frac{-\log \mathbb{P}(\xi > z)}{\log z},$$

where $\xi^+ = \max(0, \xi)$ is the positive part of ξ . Let ξ_1 and ξ_2 be random variables such that

$$(2) \quad \mathbb{I}(\xi_1) = \mathbb{I}(\xi_2) = \alpha,$$

where $\alpha \in (0, \infty)$. Assume further that ξ_1 and ξ_2 are non-negative and independent.

Show that

$$\mathbb{I}(\xi_1 \xi_2) = \alpha.$$

Exercise 1.4 (Exercise 1.3 continues). Let ξ_1 and ξ_2 be real-valued random variables that satisfy (2) and $\mathbb{P}(\xi_1 > z) = \mathbb{P}(\xi_2 > z)$ for all large enough z , but which are not necessarily independent.

Show that

$$\mathbb{I}(\max(\xi_1, \xi_2)) = \alpha.$$

Exercise 1.5 (Exercise 1.4 continues). Let ξ_1 and ξ_2 be non-negative random variables that satisfy (2) and $\mathbb{P}(\xi_1 > z) = \mathbb{P}(\xi_2 > z)$ for all large enough z , but which are not necessarily independent.

Show that

$$\mathbb{I}(\xi_1 + \xi_2) = \alpha.$$

Does the result hold if ξ_1 and ξ_2 are allowed to obtain negative values?