

# Risk Theory (NMF503) – practicals

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## 1 Point processes – Poisson process

Basic notation and properties for the homogeneous Poisson process with intensity  $\lambda$ :

- $N_t$  – random number of events observed until time  $t$  with distribution  $\text{Po}(\lambda t)$ ,
- $\sigma_n$  – random time of  $n$ -th event with Erlang distribution with pdf

$$f(x) = \frac{\lambda^n}{(n-1)!} x^{n-1} e^{-\lambda x}, \quad x \geq 0,$$

- $\tau_i$  – (independent) random time between events  $i-1$  and  $i$  with exponential distribution  $\text{Exp}(\lambda)$ .

**Example 1.1.** Construct a random sample from Poisson distribution using the properties of the Poisson process.

Hint. Let  $\{U_k\}_{k=1,2,\dots}$  be iid with uniform distribution on  $[0, 1]$ . Set

$$T_k := \prod_{i=1}^k U_i,$$

and

$$N := \inf\{n : T_n < e^{-\lambda}\}.$$

Show that  $N - 1 \sim \text{Po}(\lambda)$ .

**Example 1.2.** Let  $n$  events of homogeneous Poisson process be observed during the time period  $[0, T]$  at times  $\sigma_1 = s_1, \dots, \sigma_n = s_n$ .

- Derive the maximum-likelihood estimate of the intensity  $\lambda$ .
- Verify the properties of the ML estimate: unbiasedness, consistency.
- Derive the sufficient statistic(s) for the estimate.
- Construct a confidence interval for the intensity.

**Example 1.3.** Let  $n$  events of homogenous Poisson process with parameter  $\lambda$  be observed during the time period  $[0, T]$  at times  $\sigma_1 = s_1, \dots, \sigma_n = s_n$  and let  $n'$  events of homogenous Poisson process with parameter  $\lambda'$  be observed during the time period  $[0, T']$  at times  $\sigma'_1 = s'_1, \dots, \sigma'_{n'} = s'_{n'}$ . Let the processes be independent. Propose a statistical test of the hypothesis  $\lambda = \lambda'$ .

**Example 1.4.** Generalize the above test to  $k$  independent homogenous Poisson processes, i.e. derive a test of hypothesis  $\lambda = \lambda_2 = \dots = \lambda_k$ .

**Example 1.5.** Let  $n$  events of **nonhomogenous** Poisson process be observed during the time period  $[0, T]$  at times  $\sigma_1 = s_1, \dots, \sigma_n = s_n$ . Consider intensity

$$\lambda(t) = e^{\alpha + \beta t}.$$

Derive a statistical test of hypothesis  $\beta = \beta_0$ , and focus on the case  $\beta_0 = 0$ , i.e. construct a test of homogeneity.

## 2 Collective risk model and ruin probability, subexponential distributions

**Example 2.1.** Under the standard assumptions (compound Poisson process, costs, premium, see the Lecture notes for details) derive the adjustment coefficient  $R$  when the severity distribution follows  $\Gamma(\frac{1}{2}, \beta)$  with pdf

$$p(x) = \frac{\sqrt{\beta}}{\Gamma(\frac{1}{2})} x^{-\frac{1}{2}} e^{-\beta x}.$$

**Example 2.2.** Derive the Laplace transform of the Beekman's formula. Then compute the probability of ruin under the exponential distribution of claims severity, i.e.

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

**Example 2.3.** Verify that the distribution with pdf

$$p(x) = \frac{a}{\sqrt{2\pi}} x^{-\frac{3}{2}} e^{-\frac{a^2}{2x}}$$

belongs to the subexponential family.

**Example 2.4.** Show that the standard criterion does not show that the lognormal distribution belongs to the subexponential family.

**Example 2.5.** (\*) Derive an asymptotic formula for the ruin probability under the lognormal distribution of the claim severity.

**Example 2.6.** Consider Excess of Loss (XL) reinsurance with priority  $a > 0$  and layer  $L > 0$ . Let the claims of insurer follow the compound Poisson process. Elaborate the claims from the point of view of the reinsurer.

**Example 2.7.** Derive an estimate of the parameter of Pareto distribution based on the quantiles.

### 3 Extreme Value Theory

**Example 3.1.** Consider Fréchet distribution  $X$  with cdf

$$G_{1,\alpha}(x) = \exp(-x^{-\alpha}), \quad x > 0, \quad \alpha > 0.$$

Verify that for the moments it holds

$$\mathbb{E}[X^j] = \Gamma(1 - \frac{j}{\alpha}), \quad j < \alpha.$$

**Example 3.2.** Verify the max-stability of the extreme value distributions, i.e. that for cdf  $G$  and proper choices of the sequences  $\{c_n > 0\}$ ,  $\{d_n\}$ , it holds

$$G^n(c_n x + d_n) = G(x), \quad n = 1, 2, \dots$$

**Example 3.3.** Show that the extreme value distributions can be used to deal with the distributions of the minima of a sequence of random variables.

Hint: Show that

$$P\left(\max_{i \leq n}(-X_i) \leq a_n x + b_n\right) = 1 - P\left(\min_{i \leq n}(X_i) \leq -a_n x - b_n\right),$$

where  $a_n > 0$ ,  $b_n \in \mathbb{R}$ .

**Example 3.4.** Consider generalized Pareto distribution with cdf

$$W_{\gamma,\mu,\sigma}(x) = 1 - \left(1 + \gamma \frac{x - \mu}{\sigma}\right)^{-\frac{1}{\gamma}},$$

where  $\gamma \neq 0$ ,  $\mu \in \mathbb{R}$ ,  $\sigma > 0$ . For  $\gamma > 0$  the support is  $x \geq \mu$ , whereas for  $\gamma < 0$  we have  $\mu \leq x \leq \mu - \frac{\sigma}{\gamma}$ . Show that

$$\mathbb{E}[X] = \mu + \frac{\sigma}{1 - \gamma}, \text{ if } \gamma < 1,$$

and

$$\mathbb{E}[X] = \infty, \text{ if } \gamma \geq 1.$$

**Example 3.5.** Explore the limiting tail behaviour and derive the domain of attraction for the following distributions:

1. Pareto,
2. Exponential,
3. Beta.

**Example 3.6.** Consider a sequence of i.i.d. random variables with distribution function  $F(x)$ . Derive cdf for the maximum over random number of random variables, where the random number follows Poisson distribution with parameter  $\lambda$ . Then consider the case when  $F(x)$  corresponds to generalized Pareto distribution with parameters  $(\gamma, 0, \sigma)$ .

## 4 Copula functions

**Example 4.1.** Consider bivariate discrete distribution with realization and probabilities

$$\begin{aligned} P(X_1 = 0, X_2 = 0) &= \frac{1}{8}, & P(X_1 = 1, X_2 = 1) &= \frac{3}{8}, \\ P(X_1 = 0, X_2 = 1) &= \frac{2}{8}, & P(X_1 = 1, X_2 = 0) &= \frac{2}{8}. \end{aligned}$$

Derive the marginal distributions and discuss the (non)uniqueness of the copula function which represents the dependence.

**Example 4.2.** Consider the multivariate and bivariate Gaussian copula. Derive an explicit formula for  $\rho \in \{-1, 0, 1\}$ .

**Example 4.3.** Consider the bivariate Gumbel copula with parameter  $\theta \in [1, \infty)$ . Compute the limit for  $\theta \rightarrow \infty$ .

**Example 4.4.** Consider the bivariate Clayton copula with parameter  $\theta \in (0, \infty)$ . Compute the limit for  $\theta \rightarrow \infty$  and  $\theta \rightarrow 0_+$ .