

**Exercise 1.** Let  $\eta : \mathcal{B}(\mathbb{X}) \rightarrow \mathbb{N}$  be a Poisson point process with intensity  $\lambda$ , i.e.  $\eta(A)$  is the number of points in  $A \in \mathcal{B}(\mathbb{X})$ . Let  $0 < p \leq 1$ . Remove each point produced by the process independently with probability  $1 - p$ . Show that the resulting point process  $\eta'$  is a Poisson point process with intensity  $p\lambda$ .

**Solution:** The independence of the random variables  $\eta'(A)$  and  $\eta'(B)$  in  $A \cap B = \emptyset$  is clear from the assumptions. Hence we only need to show that the said random variables are Poisson distributed.

Write  $\lambda_A = \lambda\mu(A)$ . Then for all non-negative integers  $n$

$$\mathbb{P}[\eta(A) = n] = \frac{\lambda_A^n}{n!} e^{-\lambda_A}.$$

Hence

$$\mathbb{P}[\eta'(A) = k] = \sum_{n=k}^{\infty} \mathbb{P}[\eta(A) = n] \mathbb{P}[B_n = k].$$

where  $B_n$  are independent random variables with  $B_n \sim \text{Bin}\left(\frac{n}{p}\right)$ . Hence

$$\begin{aligned} \mathbb{P}[\eta'(A) = k] &= \sum_{n=k}^{\infty} \binom{n}{k} p^k (1-p)^{n-k} \frac{\lambda_A^n}{n!} e^{-\lambda_A} \\ &= \frac{p^k}{k!} e^{-\lambda_A} \lambda_A^k \sum_{n=k}^{\infty} \frac{(\lambda_A - p\lambda_A)^{n-k}}{(n-k)!} \\ &= \frac{p^k}{k!} e^{-\lambda_A} \lambda_A^k e^{\lambda_A - p\lambda_A} \\ &= \frac{(p\lambda_A)^k}{k!} e^{-p\lambda_A} \end{aligned}$$

**Exercise 2.** Consider a finite measure space  $X$  and a sequence of measurable sets  $A_1, A_2 \cdots \subset X$ . Show that if

$$\limsup \mu(A_n) \geq \lambda.$$

Then  $\mu(\limsup A_i) \geq \lambda$ . *Hint: Fatou's lemma.*

**Solution:** By Fatou's lemma

$$\begin{aligned} \lambda \leq \limsup \mu(A_n) &= \mu(X) - \liminf \int_X \chi_{A_n} d\mu \leq \mu(X) - \int_X \liminf \chi_{A_n} d\mu \\ &= \mu(X) - \mu(\liminf A_n) = \mu(\limsup A_n) \end{aligned}$$

**Exercise 3.** Use Karamata's representation theorem to verify the following facts for  $f \in RV_\alpha$ ,  $g \in RV_\beta$ :

- If  $\alpha > 0$  then  $f \rightarrow \infty$ , if  $\alpha < 0$ , then  $f \rightarrow 0$  as  $x \rightarrow \infty$ .
- If  $\alpha > \beta$  then  $f/g \rightarrow \infty$ .
- $f + g \in RV_{\max\{\alpha, \beta\}}$

*Proof.* Assume  $\alpha > 0$ . By Karamata's representation theorem for  $t > t_0$

$$f(t) = c(t) \exp \left( \int_{t_0}^t \frac{a(s)}{s} ds \right).$$

Then for  $t > t_\delta$  we have  $a(s) \geq \alpha/2$  and

$$\begin{aligned} f(t) &= c(t) \exp \left( \int_{t_\delta}^t \frac{a(s)}{s} ds + \int_{t_0}^{t_\delta} \frac{a(s)}{s} ds \right) \geq c(t) \exp \left( \int_{t_\delta}^t \frac{a(s)}{s} ds + O(1) \right) \\ &= \exp \left( \frac{\alpha}{2} \log t + O(1) \right) \rightarrow \infty. \end{aligned}$$

The case  $\alpha < 0$  is similar.

If  $f/g$  in  $RV_{\alpha-\beta}$ . If  $\alpha > \beta$  the result follows from the previous part.

Assume that  $\alpha > \beta$ . Note that

$$\frac{f(tx) + g(tx)}{f(t) + g(t)} = \frac{\frac{f(tx)}{f(t)} + \frac{g(tx)}{f(t)}}{1 + \frac{f(t)}{g(t)}} = \frac{f(tx)}{f(t)} + o(1) \rightarrow x^\alpha.$$

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