



## **Mathematical Statistics**

## Seminar Exercises: Week 6

**Recap.** For any random variables X, Y and  $\alpha \in \mathbb{R}$  we have:

• 
$$E(X + Y) = E(X) + E(Y)$$
;

• 
$$E(\alpha X) = \alpha E(X);$$

• 
$$V(\alpha X) = \alpha^2 V(X)$$
;

If in addition X and Y are independent, then:

• 
$$E(X \cdot Y) = E(X) \cdot E(Y)$$
;

• 
$$V(X + Y) = V(X) + V(Y)$$

If  $X \sim \mathcal{N}(\mu, \sigma)$ , then:

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \ x \in \mathbb{R}$$

If  $X \sim \mathcal{N}(\mu, \sigma)$ , then:

$$Y = \frac{X - \mu}{\sigma} \sim \mathcal{N}(0, 1)$$

$$Z = X - \alpha \sim \mathcal{N}(\mu', \sigma')$$

If X and Y are independent random variables with  $X, Y \sim \mathcal{N}(\mu, \sigma)$ , then for any  $\alpha, \beta \in \mathbb{R}$ :

$$\alpha X + \beta Y \sim \mathcal{N}(\mu', \sigma')$$

If  $X_1, X_2, \dots, X_n$  are i.i.d. random variables,  $X_i \sim \mathcal{N}(\mu, \sigma)$ , then:

$$\overline{X} = \frac{1}{n} \sum_{i=1}^{n} X_i \sim \mathcal{N}\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

**Exercise 1.** Let  $X_1, X_2, \ldots, X_n, \ldots$  be i.i.d. (independent identically distributed) random variables that follow the normal distribution,  $X \sim \mathcal{N}(\mu, \sigma)$ . Find the constant  $k_n$  such that the sampling function

$$\overline{s} = k_n \sum_{j=1}^{n} |X_j - \overline{X}|$$

verifies  $E(\overline{s}) = \sigma$ .

**Recap.** • If  $X_1, ..., X_n$  are independent identically distributed random variables with distribution given by  $X \sim \mathcal{N}(\mu, \sigma)$  and we define the statistic:

$$V := \frac{1}{\sigma^2} \sum_{i=1}^{n} (X_i - \overline{X})^2 = \frac{(n-1)s^2}{\sigma^2}$$

then

$$V \sim \chi^2(n-1)$$

• If  $X \sim \chi^2(n)$ , then:

$$f_X(x) = \frac{1}{2^{\frac{n}{2}}\Gamma(\frac{n}{2})}x^{\frac{n}{2}-1}e^{-\frac{x}{2}}, \ x > 0$$

where for every  $\alpha > 0$  we have:

$$\Gamma(\alpha) = \int_0^\infty x^{a-1} e^{-x} dx$$

- **Exercise 2.** (a) Let  $X \sim \chi^2(n)$ . Find the probability density function of the random variable  $Y = \sqrt{\frac{X}{n}}$ .
- (b) Let  $X_1, X_2, ..., X_n, ...$  be i.i.d. (independent identically distributed) random variables that follow the distribution  $X \sim \mathcal{N}(\mu, \sigma)$ . Find the constant  $k_n$  such that the sampling function

$$\overline{s} = k_n \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (X_i - \overline{X})^2}$$

verifies  $E(\overline{s}) = \sigma$ .

**Recap.** • A sequence  $(X_n)_{n\in\mathbb{N}}$  of random variables converges **almost** surely (denoted by **a.s.** and written as  $X_n \stackrel{a.s}{\to} X$ ) to a random variable X if:

$$P\left(\lim_{n\to\infty} X_n = X\right) = 1$$

• The Strong Law of Large Numbers (SLLN):

If  $(X_n)_{n\in\mathbb{N}}$  is a sequence of i.i.d. random variables with  $X_n \sim X$ , then

$$\overline{X}_n := \frac{1}{n} \sum_{i=1}^n X_i \xrightarrow{a.s.} E(X)$$

• If  $X \sim Unif[a,b]$ , then:

$$f_X(x) = \begin{cases} \frac{1}{b-a}, & \text{if } x \in [a,b] \\ 0, & \text{otherwise} \end{cases}$$

**Exercise 3.** Let  $(X_n)_{n\in\mathbb{N}}$  be a sequence of i.i.d. (independent identically distributed) random variables that follow the distribution

$$X \sim Unif[a, b]$$

where 0 < a < b, and consider the following statistics:

1. Arithmetic mean of selection:

$$a_n(X_1, \dots, X_n) := \frac{1}{n} \sum_{i=1}^n X_i$$

2. Geometric mean of selection:

$$g_n(X_1,\ldots,X_n) := \sqrt[n]{\prod_{i=1}^n X_i}$$

3. Harmonic mean of selection:

$$h_n(X_1, \dots, X_n) := \frac{n}{\sum_{i=1}^n \frac{1}{X_i}}$$

Prove that each of the above statistics converges almost surely to a constant, as  $n \to \infty$  and find these constants.

**Exercise 4** (Glivenko-Cantelli). Let X be a characteristic with cumulative distribution function F and  $X_1, \ldots, X_n$  sample variables for a random sample of size n with the sample distribution function  $\overline{F}_n$ , given by:

$$\overline{F}_n(x) = \frac{\#\left\{i \in \{1, \dots, n\} : X_i \le x\right\}}{n}$$

Prove that  $\overline{F}_n(x) \xrightarrow{a.s.} F(x)$ , as  $n \to \infty$ .

 $\underline{\text{Hint:}}$  Use the SLLN.

**Recap.** • If  $X \sim Bern(p)$ , then:

$$X \begin{pmatrix} 0 & 1 \\ 1 - p & p \end{pmatrix}$$

• If  $X \sim Geo(p)$ , then:

$$X \begin{pmatrix} k \\ p(1-p)^k \end{pmatrix}_{k \in \mathbb{N}}$$

• If  $X \sim Exp(\lambda)$ , then:

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & \text{for } x > 0 \\ 0, & \text{for } x \le 0 \end{cases}$$

**Exercise 5.** Let  $X_1, X_2, \ldots, X_n, \ldots$  be i.i.d. (independent identically distributed) random variables that follow the distribution of a given characteristic X with finite mean and variance. Consider the statistic

$$s_n := s_n(X_1, \dots, X_n) = X_1 + \dots + X_n, \ n \in \mathbb{N}^*$$

Find the distribution of  $s_n$  in each of the cases below (using induction, if you so desire) and then find the mean and the variance of this statistic.

- (a)  $X \sim Bern(p)$ , where  $p \in (0,1)$ .
- (b)  $X \sim Geo(p)$ , where  $p \in (0, 1)$ .
- (c)  $X \sim Exp(\lambda)$ , where  $\lambda > 0$ .