

Seminar W6 - 831

$X$  random variable,  $(X_n)_{n \in \mathbb{N}}$  sequence of random variables that are i.i.d. (independent and identically distributed),  $X_n \sim X$

Statistic:  $h: \mathbb{R}^n \rightarrow \mathbb{R}$  measurable

$$h = h(X_1, \dots, X_n)$$

$$\bar{x} = \frac{1}{n} \cdot (X_1 + X_2 + \dots + X_n)$$

$$\overline{x^k} = \frac{1}{n} (X_1^k + X_2^k + \dots + X_n^k)$$

$$\overline{\mu_k} = \frac{1}{n} ((X_1 - \bar{x})^k + \dots + (X_n - \bar{x})^k)$$

**Exercise 1.** Let  $X_1, X_2, \dots, X_n, \dots$  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

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

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

Sol.:  $\forall X, Y : E(X+Y) = E(X) + E(Y)$

$\forall \alpha \in \mathbb{R} : E(\alpha X) = \alpha E(X)$

$$E(\bar{s}) = E\left(k_n \cdot \sum_{j=1}^n |X_j - \bar{X}|\right) = k_n \cdot E\left(\sum_{j=1}^n |X_j - \bar{X}|\right)$$

$$\bar{X} = \frac{1}{n} \cdot \sum_{i=1}^n X_i$$

$$E(\bar{X}) = \frac{1}{n} \cdot \sum_{j=1}^n E(|X_j - \bar{X}|)$$

$$X_1, \dots, X_n \text{ identically distributed} \Rightarrow E(|X_j - \bar{X}|) = E(|X_1 - \bar{X}|) \quad \forall j$$

$$E(\bar{X}) = \frac{1}{n} \cdot n \cdot E(|X_1 - \bar{X}|)$$

$$X_1 \sim \mathcal{N}(\mu, \sigma) \quad , \quad E(X_1) = \mu$$

$$\sigma_{X_1} = \sqrt{V(X_1)} = \sigma$$

$$\text{Let } Y = X_1 - \bar{X}$$

$$X_1 \sim \mathcal{N}(\mu, \sigma)$$

$$\bar{X} \sim \mathcal{N}\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

If  $X$  and  $Y$  are independent random variables with  $X, Y \sim \mathcal{N}(\mu, \sigma)$ , then

for any  $\alpha, \beta \in \mathbb{R}$ :

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

$$Y = X_1 - \bar{X} = X_1 - \frac{1}{n} \cdot (X_1 + X_2 + \dots + X_n) =$$

$$= \frac{n-1}{n} \cdot X_1 + \frac{1}{n} X_2 + \dots + \frac{1}{n} X_n$$

$$\Rightarrow Y \sim \mathcal{N}(\mu', \sigma')$$

$$\mu' = E(Y) = E(X_1 - \bar{X}) = E(X_1) - E(\bar{X}) = \mu - \mu = 0$$

$$E(\bar{X}) = E\left(\frac{1}{n} (x_1 + \dots + x_n)\right) = \frac{1}{n} \cdot (E(x_1) + E(x_2) + \dots + E(x_n)) =$$

$$= \frac{1}{n} \cdot n \cdot \underbrace{E(x_1)}_{=\mu} = \mu$$

$$\sigma' = \sqrt{V(Y)}$$

$$V(Y) = V\left(x_1 - \bar{X}\right) = V\left(x_1 - \frac{1}{n} (x_1 + x_2 + \dots + x_n)\right) =$$

$$= V\left(\frac{n-1}{n} x_1 + \left(-\frac{1}{n}\right) x_2 + \dots + \left(-\frac{1}{n}\right) x_n\right) \stackrel{x_1, \dots, x_n \text{ independent}}{=} \\ = V\left(\frac{n-1}{n} x_1\right) + V\left(-\frac{1}{n} x_2\right) + \dots + V\left(-\frac{1}{n} x_n\right) =$$

$$= \left(\frac{n-1}{n}\right)^2 \underbrace{V(x_1)}_{=\sigma^2} + \frac{1}{n^2} \cdot \underbrace{V(x_2)}_{=\sigma^2} + \dots + \frac{1}{n^2} \underbrace{V(x_n)}_{=\sigma^2} =$$

$$= \sigma^2 \cdot \left(\left(\frac{n-1}{n}\right)^2 + (n-1) \cdot \frac{1}{n^2}\right) = \sigma^2 \cdot \frac{n-1}{n^2} (n-1+1) = \sigma^2 \cdot \frac{n-1}{n}$$

$$\Rightarrow Y \sim \mathcal{N}(\mu', \sigma') \quad \text{and} \quad \mu' = 0 \quad \sigma' = \sigma \sqrt{\frac{n-1}{n}}$$

$$\Rightarrow Y \sim \mathcal{N}\left(0, \sigma \sqrt{\frac{n-1}{n}}\right)$$

$$E(\bar{S}) = \frac{1}{k_n} \cdot n \cdot E(|Y|) \quad \text{the "probability" of gaining that money}$$

$$E(|Y|) = \int_{\mathbb{R}} \underbrace{|y|}_{\text{"the money"}} \cdot \underbrace{f_Y(y)}_{\text{gain by playing the game}} dy$$

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

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

$$Y \sim \mathcal{N}(0, \sigma\sqrt{\frac{n-1}{n}}) \Rightarrow f_Y(y) = \frac{1}{\sigma\sqrt{\frac{n-1}{n}} \cdot \sqrt{2\pi}} \cdot e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}}$$

$$E(|Y|) = \int_{\mathbb{R}} |y| \cdot f_Y(y) dy = \underbrace{\int_{-\infty}^0 (-y) \cdot f_Y(y) dy}_{y = -z} +$$

$$+ \int_0^{\infty} y f_Y(y) dy \quad \int_0^{\infty} z \cdot f_Y(-z) \cdot (-dz) =$$

$$= \int_0^{\infty} z \underbrace{f_Y(-z)}_{f_Y(z)} dz =$$

$$= \int_0^{\infty} y f_Y(y) dy$$

$$= 2 \int_0^{\infty} y f_Y(y) dy = 2 \int_0^{\infty} y \cdot \frac{1}{\sigma\sqrt{\frac{n-1}{n}} \cdot \sqrt{2\pi}} \cdot e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}} dy =$$

$$= \frac{2}{\sigma\sqrt{\frac{n-1}{n}} \cdot \sqrt{2\pi}} \cdot \int_0^{\infty} \underbrace{y \cdot e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}}}_{\quad} dy =$$

$$\left( e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}} \right)' = e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}} \cdot \frac{-2y}{2\sigma^2 \cdot \frac{n-1}{n}}$$

$$= \frac{2}{\sigma\sqrt{\frac{n-1}{n}} \cdot \sqrt{2\pi}} \cdot \underbrace{\left( e^{-\frac{y^2}{2\sigma^2 \cdot \frac{n-1}{n}}} \right) \Big|_0^{\infty}}_{=-1} \cdot \frac{2\sigma^2 \cdot \frac{n-1}{n}}{-2} =$$

$$= \frac{2\sigma^2 \cdot \frac{n-1}{n}}{\sigma\sqrt{\frac{n-1}{n}} \cdot \sqrt{2\pi}} = \sigma \cdot \sqrt{\frac{n-1}{n}} \cdot \sqrt{\frac{2}{\pi}}$$

$$E(\bar{S}) = \sigma \sqrt{\frac{n-1}{n}} \cdot \sqrt{\frac{2}{\pi}} \cdot k_{n-1} = \sigma$$

$$\Rightarrow k_n = \frac{1}{n\sqrt{\frac{n-1}{n}} \cdot \sqrt{\frac{2}{\pi}}}$$

Ex 2 :

Ex 3 :

**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

$$X \sim \text{Unif}[a, b]$$

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:

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

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  $n \rightarrow \infty$  and find these constants.

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

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

• **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

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

• If  $X \sim \text{Unif}[a, b]$ , then:

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

Sol. :

$$a_n(x_1, x_2, \dots, x_n) = \frac{1}{n} (x_1 + x_2 + \dots + x_n) \rightarrow E(X)$$

$$X \sim \text{Unif}([a, b]) \quad , \quad f_X(x) = \begin{cases} \frac{1}{b-a}, & x \in [a, b] \\ 0, & x \notin [a, b] \end{cases}$$

$$E(X) = \int_{(2)} x \cdot f_X(x) dx = \int_a^b x \cdot \frac{1}{b-a} dx = \frac{1}{b-a} \cdot \left( \frac{1}{2} x^2 \right) \Big|_a^b =$$

$$= \frac{1}{b-a} \cdot \frac{1}{2} (b^2 - a^2) = \frac{a+b}{2}$$

$$\Rightarrow a_n \xrightarrow{a.s.} \frac{a+b}{2}$$

$$g_n = \sqrt[n]{x_1 x_2 \dots x_n}$$

$$\ln g_n = \frac{1}{n} (\ln x_1 + \ln x_2 + \dots + \ln x_n)$$

$$Y_n := \ln x_n \Rightarrow Y_n \sim \ln X$$

$$\text{By the SLLN: } \frac{1}{n} (Y_1 + \dots + Y_n) \xrightarrow{a.s.} E(\ln X)$$

$$\begin{aligned} E(\ln X) &= \int_{(2)} \ln x \cdot f_X(x) dx = \int_a^b \ln x \cdot \frac{1}{b-a} dx = \\ &= \frac{1}{b-a} \cdot \int_a^b \ln x dx \\ \int_a^b \ln x dx &= \int_a^b x^{-1} dx = x \ln x \Big|_a^b - \int_a^b \underbrace{x \cdot (\ln x)'}_{1} dx = \\ &= b \ln b - a \ln a - (b-a) \end{aligned}$$

$$\begin{aligned} \Rightarrow \ln g_n &\xrightarrow{a.s.} \frac{1}{b-a} (b \ln b - a \ln a - (b-a)) \\ g_n &\xrightarrow{a.s.} e^{\frac{1}{b-a} (b \ln b - a \ln a - (b-a))} \end{aligned}$$

$$h_n = \frac{n}{\frac{1}{x_1} + \frac{1}{x_2} + \dots + \frac{1}{x_n}}$$

$$h_n = \left[ \frac{1}{n} \left( \frac{1}{x_1} + \frac{1}{x_2} + \dots + \frac{1}{x_n} \right) \right]^{-1}$$

$$\frac{1}{X_n} =: Y_n \quad \Rightarrow \quad Y_n \sim \frac{1}{X} =: Y$$

$$\frac{Y_1 + \dots + Y_n}{n} \xrightarrow{\text{a.s.}} E(Y) = E\left(\frac{1}{X}\right)$$

$$\Rightarrow \frac{1}{h_n} \xrightarrow{\text{a.s.}} E\left(\frac{1}{X}\right) \Rightarrow h_n \xrightarrow{\text{a.s.}} \frac{1}{E\left(\frac{1}{X}\right)}$$

$$\begin{aligned} E\left(\frac{1}{X}\right) &= \int_{\mathbb{R}} \frac{1}{x} \cdot f_X(x) dx = \int_a^b \frac{1}{x} \cdot \frac{1}{b-a} dx = \frac{1}{b-a} \cdot \ln x \Big|_a^b = \\ &= \frac{1}{b-a} (\ln b - \ln a) = \frac{\ln \frac{b}{a}}{b-a} \end{aligned}$$