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Ex1) (X_n) iid, $X_i \sim \mathcal{P}(1)$, $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$, $S_n = \sum_{i=1}^n X_i$ 3) 1) Selon TCL $\bar{X}_n \xrightarrow{d} \mathcal{N}(E(X_i), V(X_i)/n)$ (1)
or $E(X_i) = 1$ et $V(X_i) = 1$ alors $\bar{X}_n \xrightarrow{d} \mathcal{N}(1, 1/n)$ (1)2) a) $S_n = n \bar{X}_n$ alors $E(S_n) = n E(\bar{X}_n) = n$ (1)
et $V(S_n) = n^2 V(\bar{X}_n) = n^2 \times 1/n = n$
 $\Rightarrow S_n \xrightarrow{d} \mathcal{N}(n, n)$ (1)2) 3) $S_n \xrightarrow{d} \mathcal{N}(n, n)$ alors $Z = \frac{S_n - n}{\sqrt{n}} \xrightarrow{d} \mathcal{N}(0, 1)$ (1)

$$\lim_{n \rightarrow \infty} P(S_n \leq n) = \lim_{n \rightarrow \infty} P(S_n - n \leq 0) = \lim_{n \rightarrow \infty} P(Z \leq 0) = 1/2. (1)$$

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Ex2) 1) a) $E(X) = \lambda$ alors $\bar{X}_n = \lambda$, \bar{X}_n est l'EMM de λ . (1)

b) $L(\lambda) = \prod_{i=1}^n P(X=x_i) = \prod_{i=1}^n e^{-\lambda} \frac{\lambda^{x_i}}{x_i!}$ (1)

$$\bullet \ell(\lambda) = \log L(\lambda) = \sum_{i=1}^n \left[\ln(e^{-\lambda}) + \ln \frac{\lambda^{x_i}}{x_i!} \right]$$

$$= \sum_{i=1}^n \left[-\lambda + x_i \ln \lambda - \ln x_i! \right] (1)$$

$$\bullet \frac{d\ell(\lambda)}{d\lambda} = \sum_{i=1}^n \left[-1 + x_i/\lambda \right] = 0 \text{ alors } -n + \frac{\sum x_i}{\lambda} = 0 \quad (1)$$
$$\Rightarrow \frac{\sum x_i}{\lambda} = n \Rightarrow \lambda = \frac{1}{n} \sum x_i = \bar{X}_n.$$
$$\Rightarrow \bar{X}_n \text{ est l'EMV de } \lambda.$$

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2) Soit $f(x) = (\theta + 1)x^\theta \times \mathbb{1}_{[0,1]}$

$$\bullet L(\theta) = \prod_{i=1}^n (\theta + 1) x_i^\theta \quad 0.5$$

$$\bullet \ell(\theta) = \sum_{i=1}^n \left[\ln(\theta + 1) + \theta \ln(x_i) \right] (1)$$

$$\bullet \frac{d\ell(\theta)}{d\theta} = \sum_{i=1}^n \left[\frac{1}{1+\theta} + \ln x_i \right] = 0 \text{ alors } \frac{n}{1+\theta} = -\sum_{i=1}^n \ln x_i$$

$$\text{et } 1+\theta = -\frac{n}{\sum \ln x_i} \Rightarrow \theta = 1 + \frac{n}{\sum \ln x_i} \text{ l'EMV de } \theta. (1)$$

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Ex3

2)

1) * $E(\theta_1) = E\left(\frac{x_1 + x_2}{2}\right) = \frac{1}{2}(\mu + \mu) = \mu \Rightarrow$ non biaisé

* $E(\theta_2) = E\left(\frac{x_1 + 3x_2}{4}\right) = \frac{1}{4}\mu + \frac{3}{4}\mu = \mu \Rightarrow$ non biaisé

2) Non biaisés alors on compare leur variances.

* $V(\theta_1) = \frac{1}{4}[V(x_1) + V(x_2)] = \frac{\sigma^2}{2}$. ↗ car x_1 et x_2 sont indép.

* $V(\theta_2) = \frac{1}{16}V(x_1) + \frac{9}{16}V(x_2) = \frac{10}{16}\sigma^2 = \frac{5}{8}\sigma^2$

$V(\theta_2) > V(\theta_1)$ alors on choisit θ_1 .