Sérien 2: Estimation

Exo1:

$$\begin{array}{lll}
\sqrt{1} & \chi_{1}, \chi_{2}, - , \chi_{n} & \text{independentles} \\
\theta = (\alpha_{n}, -)\alpha_{n}, \gamma_{1}, \tau^{2}), \quad \theta' = (\alpha_{n}', -)\alpha_{n}', \gamma', \tau^{2}) \\
P_{\theta} = P_{\theta'} & \Rightarrow D & (\alpha_{n}+\gamma) = \alpha_{n}'+\gamma & \Rightarrow D & (\sum_{n=1}^{\infty} \alpha_{n} + n\gamma) = \sum_{n=1}^{\infty} \alpha_{n}' + n\gamma' \\
\alpha_{n}+\gamma = \alpha_{n}'+\gamma & \Rightarrow C & (\alpha_{n}+\gamma) = C &$$

D'on
$$V = V'$$
. Ainsi le modèle est identi fiable $\nabla_{z} = \nabla_{z}'$

2)
$$\times \sim N(N_2, T^2)$$

 $Y \sim N(N_2, T^2)$
 $Y \sim N(N_2, T^2)$

X et y sont indépendantes

D'on le modèle est non identifiable.

On a
$$N_{n,j} = N + \lambda_{j} + \lambda_{n}$$
 $\theta = (\lambda_{n} -) \lambda_{n}, \lambda_{n} - (\lambda_{n} - \lambda_{n}, \lambda_{n}, \nabla_{n})$
 $\theta_{n} = (0, -) 0, \lambda_{n} - (\lambda_{n} - \lambda_{n}) \nabla_{n} \nabla_{n}$
 $\theta_{n} = (\lambda_{n} -) \lambda_{n}, \lambda_{n} - (\lambda_{n} - \lambda_{n}) \nabla_{n} \nabla_{n}$
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Ainsi le modèle est non identifiable.

2) Prenons
$$z = (1, -, 1)$$
, $y = (\frac{1}{2}, \frac{1}{2}, 2, 1, -, 1)$

$$\sum_{i=1}^{n} z_{i} = 1$$

$$\sum_{i=1}^{n} y_i = n$$

A Complete the Com mais Žln (xi) =0 et Žln(gi) = -ln2.

Ainsi Tret Te ne sont pas équivalentes.

$$T_{\lambda}(\underline{x}) = \left(\underbrace{\sum_{\lambda = 1}^{n} x_{i}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} x_{i}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} x_{i}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} y_{i}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} y_{i}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} (y_{i} - y_{i})^{2}}_{\lambda = 1}^{n} \underbrace{\sum_{\lambda = 1}^{n} (y_{i} - y_{i})^{2}}_{\lambda$$

Acnosi T, (3)=T,(4).

Eau3:

Les Cas: Calcul direct:

$$P[X=x]=\frac{e^{\Theta}}{x!}\int_{N}^{\infty}|x|$$

Der une moyenne de Poisson

M'est pous une poisson

Somme d'une poisson est

une poisson

une poisson

$$P\left[X=z \mid T(x)=t\right] = \frac{P\left[X=z, T(x)=t\right]}{T(x)=t} (x)$$

$$P[T(X)=t] = e^{n\theta} \frac{(n\theta)^{t}}{t!} I_{IN}(t)$$

$$(*) = \int \frac{P[X=x]}{P[T(x)=t]} \sin \frac{x}{x^{t}} z_{t} = t \qquad P[X=x] = \prod_{i=1}^{n} e^{i\theta} \frac{\theta^{i}}{x!} I_{IN}(x_{i})$$

$$= e^{n\theta} \frac{x^{n}}{x^{n}} x_{i}$$

$$= e^$$

$$(x) = \int \frac{e^{n\theta} \int_{|z|}^{2\pi i} z_{i}}{\prod_{i=1}^{n} (x_{i})!} \int_{|z|}^{2\pi i} \frac{t!}{n!} \int_{|z|}^{2\pi i} z_{i} = t$$
Sinon

$$= \begin{cases} \frac{t!}{n} \frac{1}{|x|} \frac{1}{|x|} \\ \frac{1}{|x|} \frac{1}{|x|} \frac{1}{|x|} \frac{1}{|x|} \end{cases}$$
 si $\sum_{i=1}^{n} x_i = t$

Ainsi Test exhaustive.

2 eme (as: Théorème de factorisation:
$$\mathcal{L}(z, \theta) = \frac{e^{n\theta}}{\pi} \frac{2\pi i}{(\pi i)!} \int_{N^n} (z)$$

Montrons que
$$\phi = 0$$
.

E[$\phi(T(X))$] = $\sum_{k=0}^{\infty} \phi(k) e^{in\theta} \frac{(n\theta)^k}{N} (k) = 0$

$$\frac{1}{k!} \left(\frac{1}{k!} \right) = \sum_{k=0}^{\infty} \phi(k) \frac{1}{k!} \frac{$$

Il s'agrit d'une sais entière ayant pour terme général nul. Ainsi $\phi \equiv 0$.

Don Test complete.

Variante de 1) " les Cas"

On reprend de
$$P[X=X] = \frac{e^{n\theta}}{|x|} \frac{\sum_{i=1}^{n} x_i}{|x_i|}$$

$$P[X=X] T(X)=t] = P[X_i = x_i] \longrightarrow X_i = x_i \longrightarrow \sum_{i=1}^{n} x_i = t$$

$$P[T(X)=t] = P[X_i = x_i] \longrightarrow X_i = t \longrightarrow \sum_{i=1}^{n-1} x_i$$

$$P[X_i = x_i] \longrightarrow X_i = t \longrightarrow \sum_{i=1}^{n-1} x_i$$

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$$\frac{\exists x \circ 8}{b / \mathbb{P}[x = x]} = \frac{\theta^{x} e^{\theta}}{x!}$$

$$\frac{1}{x!} = \frac{\theta^{x} e^{\theta}}{x!} = \frac{1}{n^{\theta}} = \frac$$

=DT est exhaustive =DT est complète

Exo4:

1)
$$f(x,\theta) = \theta x^{\theta-1}$$

$$\chi(x,\theta) = \theta^{n} \prod_{i=1}^{n} x_{i}^{\theta-1} \text{ II } (x)$$

$$= \exp(n \log \theta + (\theta-1) \log | \prod_{i=1}^{n} x_{i}^{\theta-1}) \text{ II } (n)$$

$$= \exp(n \log \theta + \theta \log \prod_{i=1}^{n} - \log \prod_{i=1}^{n}) \text{ II } (n)$$

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$$= \exp(n \log \theta + \theta \log \prod_{i=1}^{n} - \log \prod_{i=1}^{n} \log n) \text{ II } (n)$$

$$= \exp(n \log \theta + \theta \log \prod_{i=1}^{n} - \log n) \text{ II } (n)$$

T(z) = Žlogzi est une statistique exhoustive.

$$T(z) = \sum_{i=1}^{n} z_i^{\alpha}$$
 est une statistique descriptive.

$$\chi(\bar{x},\theta) = \left(\theta^{n} \alpha^{n} / \prod_{i=1}^{n} \alpha_{i}^{\theta+i}\right) \Lambda (\bar{x})$$

$$(30,+\infty)^{n}$$

$$= \frac{\partial^2 a^{\theta}}{\partial a^{\theta}} \int_{a=1}^{\infty} da da = 0$$

$$= \frac{\partial^2 a^{\theta}}{\partial a^{\theta}} \int_{a=1}^{\infty} da = 0$$

6.

$$X(3,0) = \exp\left(n\log \theta + n\theta \log \alpha - 0 \stackrel{?}{\underset{k=1}{\sum}} \ln x_{k}\right) \frac{1}{12\alpha_{k+1}} \frac{1}{12\alpha_{k+1}}$$

$$T(x) = \stackrel{?}{\underset{k=1}{\sum}} \ln x_{k}$$

$$T(x) = \frac{1}{2} \ln x_{k} = \frac{1}{12\alpha_{k+1}}$$

$$\frac{1}{2} \times \frac{1}{2} \times \frac{1}{2} \ln x_{k} = \frac{1}{12\alpha_{k+1}}$$

$$\frac{1}{2} \times \frac{1}{2} \times \frac{$$

$$= \frac{1}{|I|} \quad \emptyset \stackrel{\sum}{=} \frac{1}{|I|} \{x_i = v_j\} \quad \lambda_1 \{x_j\}$$

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$$= \frac{1}{|I|} \quad \emptyset \stackrel{\sum}{=} \frac{1}{|I|} \{v_j\} \quad \lambda_2 \{v_j\} \quad \lambda_3 \{v_j\} \quad \lambda_4 \{v_j\} \quad$$

=
$$\int \frac{\eta_1! \times \dots \times \eta_{k+1}!}{\eta_1!} d\eta_2 v_{i_1} - yv_{k+1}$$
 indépendant de θ

Si'non

=DN est une statistique exhaustive.