

Conditional Independence between Markov Chains (Simulations) ¹

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1 Criterion of the conditional independence

The aim of this work is to develop a methodology for investigating the conditional dependence between Markov chains, that is, the objective is to obtain a procedure to verify if the process are conditionally dependents. This procedure is based on the Bayesian information criterion (BIC) [1, 2].

Consider the finite order Markov stationary process, $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$, with same alphabet \mathcal{A} . Moreover, consider that $(Z_t)_{t \in \mathbb{Z}} = (X_t, Y_t)_{t \in \mathbb{Z}}$ is an finite order Markov stationary process. Consider $r \in \mathcal{A}^k$ and $s \in \mathcal{A}^k$ fixed strings.

Definição 1. *Consider $r \in \mathcal{A}^k$ and $s \in \mathcal{A}^k$ fixed strings. The k order Markov stationary process, $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$, are conditionally independents given (r, s) if*

$$\begin{aligned} \text{Prob}(X_{t+1} = a \quad , \quad Y_{t+1} = b | X_{t-k+1}^t = r, Y_{t-k+1}^t = s) = \\ \text{Prob}(X_{t+1} = a | X_{t-k+1}^t = r) \text{Prob}(Y_{t+1} = b | Y_{t-k+1}^t = s) \end{aligned}$$

for any $a, b \in \mathcal{A}$ and for any t .

Let o_{Z^1} the order of the process $(Z_t)_{t \in \mathbb{Z}}$ when $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$ aren't conditionally independents given (r, s) and o_{Z^0} the order of the process $(Z_t)_{t \in \mathbb{Z}}$ when $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$ are conditionally independents given (r, s) . Consider $k = \max \{o_{Z^1}, o_{Z^0}\}$.

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Let $x_1^n = x_1 \dots x_{n-1} \dots x_n$, $y_1^n = y_1 \dots y_{n-1} \dots y_n$ e $z_1^n = (x_1, y_1) \dots (x_n, y_n)$ samples, respectively, of the process $(X_t)_{t \in \mathbb{Z}}$, $(Y_t)_{t \in \mathbb{Z}}$ and $(Z_t)_{t \in \mathbb{Z}}$.

The BIC of process $(Z_t)_{t \in \mathbb{Z}}$, when the process $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$ are conditionally independents, is

$$\begin{aligned} BIC^o(z_1^n) = & - \sum_{i=1}^k \sum_{a=1}^k N^X(i, a) \log \left(\frac{N^X(i, a)}{N^X(i)} \right) - \sum_{j=1}^k \sum_{b=1}^k N^Y(j, b) \log \left(\frac{N^Y(j, b)}{N^Y(j)} \right) \\ & + \frac{(|\mathcal{A}| - 1)|\mathcal{A}|^k}{2} \log(n) + \frac{(|\mathcal{A}| - 1)|\mathcal{A}|^k}{2} \log(n) \end{aligned} \quad (1.1)$$

where $N^X(i, a)$ is the number of occurrences of the string $i \in \mathcal{A}^k$ followed by the letter $a \in \mathcal{A}$ in the sample x_1^n and $N^Y(j, b)$ is the number of occurrences of the string $j \in \mathcal{A}^k$ followed by the letter $b \in \mathcal{A}$ in the sample y_1^n , such that, $N^X(i, a) = |\{k < m \leq n : x_{m-k}^{m-1} = i, x_m = a\}|$ and $N^Y(j, b) = |\{k < m \leq n : y_{m-k}^{m-1} = j, y_m = b\}|$.

The BIC of process $(Z_t)_{t \in \mathbb{Z}}$, when the process $(X_t)_{t \in \mathbb{Z}}$ and $(Y_t)_{t \in \mathbb{Z}}$ are conditionally dependents given (r, s) , is

$$\begin{aligned} BIC^o((r, s), z_1^n) = & - \sum_{\substack{i, j \in \mathcal{A}^k \\ a, b \in \mathcal{A} \\ i \neq r \\ j \neq s}} N^{XY}((i, a), (j, b)) \log \left(\frac{N^{XY}((i, a), (j, b))}{N^{XY}(i, j)} \right) \\ & - \sum_{a, b \in \mathcal{A}} N^{XY}((r, a), (s, b)) \log \left(\frac{N^X(r, a)}{N^X(r)} \right) - \sum_{a, b \in \mathcal{A}} N^{XY}((r, a), (s, b)) \log \left(\frac{N^Y(s, b)}{N^Y(s)} \right) \\ & + \frac{(|\mathcal{A}|^2 - 1)(|\mathcal{A}|^{2k} - 1)}{2} \log(n) + \frac{|\mathcal{A}| - 1}{2} \log(n) + \frac{|\mathcal{A}| - 1}{2} \log(n) \end{aligned} \quad (1.2)$$

where $N^{XY}((i, a), (j, b))$ is the number of occurrences of the string $i \in \mathcal{A}^k$ followed by the letter $a \in \mathcal{A}$ in the sample x_1^n and of the string $j \in \mathcal{A}^k$ followed by the letter $b \in \mathcal{A}$ in the sample y_1^n , such that, $N^{XY}((i, a), (j, b)) = |\{k < m \leq n : (x_{m-k}^{m-1} = i, y_{m-k}^{m-1} = j), (x_m = a, y_m = b)\}|$

The proposed procedure establishes that

$$\lim_{n \rightarrow \infty} I_{\{BIC^1(z_1^n) - BIC^o((r, s), z_1^n) \geq 0\}} = 1$$

if and only if

$$P_{(ra)(sb)}^{XY} = P_{ra}^X P_{sb}^Y$$

for any $a, b \in \mathcal{A}$ and $i, j \in \mathcal{A}^k$ where $P_{ra}^X = Prob(X_{t+1} = a | X_{t-k}^t = r)$, $P_{sb}^Y = Prob(Y_{t+1} = b | Y_{t-k}^t = s)$, $P_{(ra)(sb)}^{XY} = Prob((X_{t+1} = a, Y_{t+1} = b) | (X_{t-k}^t = r, Y_{t-k}^t = s))$, $P_{(ra)(sb)}^{XY} = Prob(X_{t+1} = a, Y_{t+1} = b | X_{t-k}^t = r, Y_{t-k}^t = s)$ and $I(\cdot)$ is the characteristic function.

We show the proof of the relation above and show three simulation scenarios.

2 Simulations

2.1 Simulation 1:

Consider the Markov stationary process of order 1, $(Z_t)_{t \in \mathbb{Z}} = (X_t, Y_t)_{t \in \mathbb{Z}}$, with alphabet $\{0, 1, 2, 3\}$. The transition probabilities $P_{(ra)(sb)}^{XY}, \forall a, b, r, s \in \{0, 1, 2, 3\}$ are shown by matrix (2.3) where their rows and columns represent, respectively, r (ou s) e a (ou b).

$$T^Z = \begin{pmatrix} 0,25 & 0,25 & 0,25 & 0,25 \\ 0,2 & 0,3 & 0,2 & 0,3 \\ 0,2 & 0,2 & 0,3 & 0,3 \\ 0,16 & 0,24 & 0,24 & 0,36 \end{pmatrix} \quad (2.3)$$

We simulated 500 samples, the length 3.000, 10.000, 15.000 e 50.000, of the process $(Z_t)_{t \in \mathbb{Z}}$.

2.2 Simulation 2:

Consider the Markov stationary process of order 2, $(Z_t)_{t \in \mathbb{Z}} = (X_t, Y_t)_{t \in \mathbb{Z}}$, with alphabet $\{0, 1, 2, 3\}$. The transition probabilities $P_{(ra)(sb)}^{XY}, \forall a, b \in \{0, 1, 2, 3\}$,

$\forall r, s \in \{00, 10, 20, 30, 01, 11, 21, 31, 02, 12, 22, 32, 03, 13, 23, 33\}$, are shown by matrix (2.4).

$$T^Z = \begin{pmatrix} 0,36 & 0,24 & 0,24 & 0,16 \\ 0,18 & 0,42 & 0,12 & 0,28 \\ 0,18 & 0,12 & 0,42 & 0,28 \\ 0,09 & 0,21 & 0,21 & 0,49 \\ 0,18 & 0,42 & 0,12 & 0,28 \\ 0,43 & 0,18 & 0,28 & 0,11 \\ 0,09 & 0,21 & 0,21 & 0,49 \\ 0,21 & 0,09 & 0,49 & 0,21 \\ 0,24 & 0,16 & 0,36 & 0,24 \\ 0,12 & 0,28 & 0,18 & 0,42 \\ 0,48 & 0,32 & 0,12 & 0,08 \\ 0,24 & 0,56 & 0,06 & 0,14 \\ 0,12 & 0,28 & 0,18 & 0,42 \\ 0,28 & 0,12 & 0,42 & 0,18 \\ 0,24 & 0,56 & 0,06 & 0,14 \\ 0,56 & 0,24 & 0,14 & 0,06 \end{pmatrix} \quad (2.4)$$

We simulated 500 samples, the length 3.000, 10.000, 15.000 e 50.000, of the process $(Z_t)_{t \in \mathbb{Z}}$.

2.3 Simulation 3:

Consider the Markov stationary process of order 1, $(Z_t)_{t \in \mathbb{Z}} = (X_t, Y_t)_{t \in \mathbb{Z}}$, with alphabet $\{0, 1, 2, 3\}$. The transition probabilities $P_{(ra)(sb)}^{XY}, \forall a, b, r, s \in \{0, 1, 2, 3\}$, are shown by matrix (2.5).

$$T^Z = \begin{pmatrix} 0,25 & 0,25 & 0,25 & 0,25 \\ 0,5 & 0 & 0 & 0,5 \\ 0,25 & 0,25 & 0,25 & 0,25 \\ 0,5 & 0 & 0 & 0,5 \end{pmatrix} \quad (2.5)$$

We simulated 500 samples, the length 3.000, 10.000, 15.000 e 50.000, of the process $(Z_t)_{t \in \mathbb{Z}}$.

Referências

- [1] CSISZÁR, I., AND TALATA, Z. Context tree estimation for not necessarily finite memory processes, via bic and mdl. *IEEE Transactions on Information Theory* 52, 3 (2006), 1007–1016.
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