Applications of Chatterjee's Correlation in MCMC (UGP, 21-22 Even Semester)

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Markov Chain Monte Carlo

Definition

A Markov chain is a discrete time stochastic process X_1, X_2, \ldots such that the next state of the process depends solely on the present state, i.e.

$$\Pr(X_{n+1}|X_1,\ldots,X_n)=\Pr(X_{n+1}|X_n).$$

- A Markov chain can be specified by two things
 - **1** The initial distribution, i.e. the marginal of X_1 .
 - 2 The transition probabilites, i.e. the conditional distribution of $X_{n+1}|X_n$

Definition

A Markov transition kernel is a map $P: \mathbb{R} imes \mathcal{B}(\mathbb{R}) o [0,1]$ such that

- for all $A \in \mathcal{B}(\mathbb{R})$, P(.,A) is a measurable function of R.
- for all $x \in \mathbb{R}$, P(x, .) is a probability measure of $\mathcal{B}(\mathbb{R})$.



Markov chain Monte Carlo

Definition

A Markov chain is said to be **stationary** if the marginal of X_n is independent of n. This invariant distribution is called the stationary distribution.

Definition

A Markov chain is **ergodic** if the distribution of X_n converges to the invariant distribution.

Definition

The Markov transition kernel P is F-symmetric if

$$F(dx)P(x,dy) = F(dy)P(y,dx).$$

This equation is often called the **detailed balance**, and the phenomenon is called *F*-reversibility

Pearson's Correlation Coefficient

 Pearson's Correlation Coefficient measures the linear correlation between two sets of data.

Definition

Given random variables X, Y, Pearson Correlation ρ is defined as

$$\rho = \frac{\mathsf{Cov}(X,Y)}{\sqrt{\mathsf{Var}(X)} \cdot \sqrt{\mathsf{Var}(Y)}}.$$

 Pearson Correlation is very powerful in detecting monotone relations and has a well developed asymptotic theory.

Problems with Pearson Correlation

There are some problems with the Pearson Correlation.

- We would like the correlation to be close to its maximum value iff one variable is a function of the other. In Pearson's case, it is equal to ± 1 iff the variables are linearly dependent.
- We would also like the correlation to be 0 iff the variables are independent. If the variables are independent, Pearson correlation is indeed 0, but the converse is not always true.



Chatterjee's Correlation Coefficient

 In order to solve these problems, Chatterjee came up with a new measure of dependence in [Chatterjee, 2020]. It overcomes the above mentioned drawbacks, and has a computationally efficient and consistent estimator.

Definition

Given random variables X, Y, where is Y is not a constant, Chatterjee correlation ξ is defined as

$$\xi(X,Y) = \frac{\int Var(\mathbb{E}(1_{\{Y \geq t\}}|X))d\mu(t)}{\int Var(1_{\{Y \geq t\}})d\mu(t)},$$

where μ is the law of Y.



Consistent Estimator of ξ

• Let $\{(X_i,Y_i)\}_{i=1}^n$ be i.i.d. pairs following the same distribution as (X,Y). Rearrange the data as $(X_{(1)},Y_{(1)}),\ldots,(X_{(n)},Y_{(n)})$ such that $X_{(1)}<\cdots< X_{(n)}$. Let r_i be the rank of $Y_{(i)}$, i.e. the number of j such that $Y_{(j)}\leq Y_{(i)}$. Then the correlation coefficient ξ_n is defined to be

$$\xi_n(X,Y) := 1 - \frac{3\sum_{i=1}^{n-1}|r_{i+1}-r_i|}{n^2-1}.$$

Theorem

If Y is not almost surely a constant, then as $n \to \infty$, $\xi_n(X,Y)$ converges almost surely to $\xi(X,Y)$.



Properties of ξ

- $\xi(X,Y) \in [0,1]$
- $\xi(X, Y) = 0$ if and only if X and Y are independent.
- $\xi(X, Y) = 1$ if and only if atleast one of X and Y is a measurable function of the other.
- ξ is not symmetric in X, Y. This is intentional and useful as we might want to study if Y is a measurable function of X, or X is a measurable function of Y. To get a symmetric coefficient, it suffices to consider $\max(\xi(X,Y),\xi(Y,X))$.
- ξ_n is based on ranks, and for the same reason, it can be computed in $O(n \log n)$.



Chatterjee's Autocorrelation Function

 We present a new autocorrelation function (ACF) using Chatterjee's correlation coefficient.

Definition

Let X_1, X_2, \ldots be a stationary, time homogeneous Markov chain with stationary distribution π . We define the new lag-k ACF as follows

$$\gamma'_{k,n} := \xi(X_n, X_{n+k}).$$



$\overline{\gamma'_{k,n}}$ is independent of n

Theorem

 $\gamma_k = Cov(X_n, X_{n+k})$ is independent of n.

Theorem

 $\gamma'_{k,n} = \xi(X_n, X_{n+k})$ is independent of n, where n and k are in \mathbb{N} .

 \bullet From now on, we'll denote $\gamma_{k,n}'$ by γ_k'



Symmetricity of γ'_k

Theorem

 $\xi(X_n, X_{n+k}) = \xi(X_{n+k}, X_n)$ for time reversal Markov chains for any $n, k \in \mathbb{N}$.



Convergence of γ_k'

Theorem

 $\lim_{n\to\infty} \xi(X_1,X_n) = 0$ for an Ergodic Markov chain.



Proof of consistency of estimator

• Chatterjee presented the proof of consistency of the estimator in [Chatterjee, 2020], where the samples drawn from (X, Y) are i.i.d.

- In our case of stationary Markov chains, we have correlated but identically distributed draws.
- We aim to prove the consistency of the estimator for our case as well.



Proof of consistency of estimator

Theorem

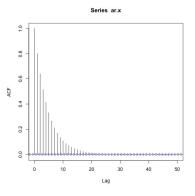
Let X_1, X_2, \ldots be a stationary time-homogeneous Markov chain with stationary distribution μ . Then $\xi_n(X,Y)$ estimated using the draws from the Markov chain converge to $\xi(X,Y)$ as $n \to \infty$, where X, Y are any two time points in the chain.

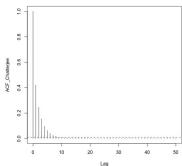
 We presented some ideas and a pathway for proving it but could not complete it and is left as future work.



Simulations

• AR(1) process with $\rho = 0.8$.

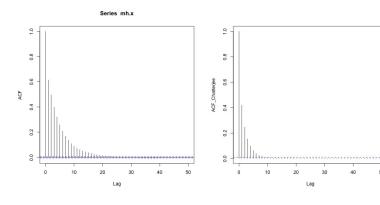






Simulations

• Metropolis-Hastings algorithm with initial distribution as Exp(0.01) and target distribution $\mathcal{N}(0,1)$.



Thank you



References I



Chatterjee, S. (2020).

A new coefficient of correlation.

