Markov chain Monte Carlo over Model Structures

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Introduction MCMCMS implements a generic framework for constructing Markov chains. It can be used to perform statistical machine learning in a Bayesian framework. It presents a modular, high level approach to performing MCMC simulations over statistical models that can explain observed data [1]. The two main benefits of MCMCMS are its prior-centric construction of the chain, that does away with the need for an explicit proposal, and the fact that complex, crisp and probabilistic, information can be encoded in the prior by use of a high level language. MCMCMS priors can be encoded as SLPs (Stochastic logic programs, [4]) or as DLPs (Distributional). DLPs extend SLPs by allowing on-the-fly computations of the probabilistic labels.

Implementation The system is implemented in Prolog and it is open source. Through extensive use of term expansion the labelled clauses are transformed to Prolog ones with additional arguments. Inserted calls to the random number generator allow probabilistic clause selection, while stochastic backtracking is via an extra clausal argument holding a path structure. MCMCMS runs on the current Yap (6.2.0) and Swi (5.11.22) Prologs.

Applications It has be tested on two model spaces: C&ARTs and Bayes nets. In addition to a number of synthetic and literature test sets, it has been used to analyse chemical binding data [5] (C&RTs) and DNA assay data [3] (BNs). [2] presented an extensive comparison on effective priors for BNs.

Availability: http://scibsfs.bch.ed.ac.uk/~nicos/sware/dlp/mcmcms

References

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