Markov chain Monte Carlo and Perfect Simulation Lecture at Aristotle University of Thessaloniki

Wilfrid S Kendall

Department of Statistics, University of Warwick





Introduction

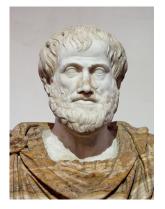


Figure 1: Αριστοτέλης 384–322 BCE

Aristotle:

- "Pleasure in the job puts perfection in the work."
- "The more you know, the more you know you don't know."

Handout available on the web: either use the QR-code



or visit https://wilfridskendall.github.io/talks/Thessaloniki-2024/.

Sketch of MCMC (I)



Figure 2: Edward Teller (1908-2003)

The original Markov chain Monte Carlo method (MCMC) was introduced by Metropolis *et al.* (1953). The senior author was Edward Teller ("father of the H-bomb").

[Fermi once said that] Teller was the only monomaniac he knew who had several manias: see Brown & May (2004).

Sketch of MCMC (II)

- Markov chain basics:
 - Transition probabilities p(a,b) (or transition rates in continuous time: unified view using exponential distribution);
 - ► Equilibrium probabilities $\pi(a)$, balance, detailed balance $\pi(a)p(a,b) = \pi(b)p(b,a)$, reversibility;
 - ▶ Aperiodicity, recurrence (and uniform and geometric recurrence);
 - under detailed balance we can condition by forbidding transitions;
- We can modify any chain, transition probabilities p(a,b), to leave a specified target distribution $\pi(a)$ invariant, by censoring each possible transition $a \to b$ with probability $\alpha(a,b) \in [0,1]$ such that $\frac{\alpha(a,b)\pi(a)p(a,b)}{\alpha(a,b)\pi(a)p(a,b)} = \frac{\alpha(b,a)\pi(b)p(b,a)}{\alpha(a,b)\pi(a)}$;
- Common choice: Metropolis-Hastings $\alpha(a,b) = \min\{1, (\pi(b)p(b,a))/(\pi(a)p(a,b))\}.$
- If result still irreducible aperiodic, then $\pi(a)$ is its long-term equilibrium.
- This is MCMC, now of intense interest to statisticians.
- But, physicists always remind us, physicists got there fifty years earlier!



Sketch of MCMC (III)

Given the $\pi(a)$, how to design a Markov chain to have this as equilibrium?

- Independence sampler: draw from a fixed probability distribution, apply Metropolis-Hastings;
- Random walk Metropolis or RWM: propose move using a random walk, apply Metropolis-Hastings;
- **Metropolis-adjusted Langevin** or MALA: make a Gaussian jump shifted using gradient of $\log \pi$, apply Metropolis-Hastings.

Can mix-and-match! RWM is often favourite: flexible, not too complicated.

Issues:

- Burn-in: How long till approximate equilibrium?
- Scaling: How big should be the RWM jump?

Question (B) is about how to get fast mixing. There is a beautiful and useful theory, but that is for another day.

Question (A) is what this lecture is all about.



Sketch of MCMC (IV)

- MCMC practicalities: Burn-in: what to do about it?
 - ▶ Theory tends to be much too pessimistic. Example: Zanella (2015a, 2015b) developed statistical methods for Anglo-Saxon history: a simplified model appeared to converge approximately in 10⁵ steps (about 1 week on compute cluster), versus 10⁹ steps in theory (around 2 centuries);
 - ▶ Is (a) one long run better or (b) many short runs? (Option (b) requires starts of short runs spread "evenly" over the sample space almost as hard in high dimensions as the original problem!)
 - ▶ Diagnostics? (Meta-theorem: for any diagnostic technique there is a chain for which the technique is deceptive!)
 - ➤ Conclusion: effective MCMC requires very careful thought about appropriate length of run think deeply about the problem!
- Can there ever be a better way?

Perfect Simulation

- The idea of exact simulation / Coupling from the Past (CFTP) / perfect simulation Propp & Wilson (1996) (Persi Diaconis: "Like seeing the landscape of Mars for the first time");
- Ideas (of "classic CFTP"):
 - extending simulation backwards through time,
 - exploit monotonicity by coupling maximal and minimal processes,
 - seek coalescence;
- Details for *random-walk-CFTP*, which can be boosted as above to provide simple image reconstruction of an image using Ising model,
 Propp & Wilson (1996) show how to vary a clever algorithm to get exact samples for critical Ising model (this is what impressed Diaconis);
- "Perfect simulation" (WSK, 1998): because everyone knows it isn't going to be perfect, whereas people might imagine "exact simulation" would somehow miraculously defeat numerical approximation error :-).

An example and some theory

- An intensely visual example, which helps many people see intuitively
 what is going on here, is *DeadLeaves-CFTP* (WSK & Thönnes, 1999)
 (technically, *Occluded CFTP*);
- What about cases where monotonicity fails? or there isn't a sensible "maximal" process? WSK (1998):
 - ▶ cross-couple upper and lower envelope processes,
 - ▶ dominate by amenable "dominating process" (time-reversible, can draw from equilibrium, can couple target processes below dominating process);
- Theoretical limits: in principle
 - Classical CFTP equivalent to uniform ergodicity (Foss & Tweedie, 1998).
 - ▶ *Dominated CFTP* is achievable under geometric ergodicity (WSK, 2004).
 - ▶ It is even possible to carry out Dominated CFTP in some **non**-geometrically ergodicity cases [Connor & WSK (2007); *nb* corrigendum];
- We can use *Dominated CFTP* to carry out perfect simulation for stable point processes (WSK & Møller, 2000);
- Detailed expositions are given by WSK (2005), Huber (2015). WSK (2015) shows how to implement CFTP in R.

Applications to Queues and Epidemics

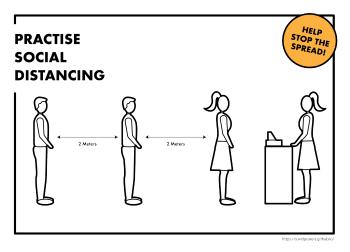


Figure 3: An illustration introducing both queues and epidemics!

Perfect Queues

The simplest queuing model (Poisson arrivals, exponential service times, single server) can be analyzed very thoroughly;

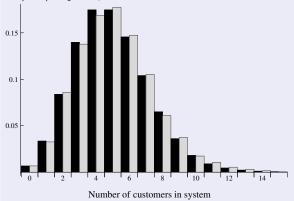
- Poisson arrivals are not unreasonable, but exponential service times are ludicrous. The M/G/1 case of general service time for just one server can use the "embedded chain" (sample at instants of departure);
- Multi-server case: computation of eg waiting-time distribution is out of reach so use simulation (and insights from Kiefer & Wolfowitz, 1955);
- Sigman (2011) shows how to do CFTP in the "super-stable" case (traffic so low that it could have been handled by just one server), using Dominated CFTP and comparing to a "Processor-sharing" discipline.

- Connor & WSK (2015) show how to extend Sigman (2011), showing how Dominated CFTP can be applied to simulate (sub-critical!) queues perfectly (and this has now been generalized by others to the case of non-Poissonian inter-arrival times). (Technical point: pathwise domination requires service times to be assigned in order of commencement of service!)
 - ▶ dominate M/G/cFCFS (FCFS means first come first served) by $M/G/cRA = [M/G/1RA]^c$ (RA means assign to individual servers on arrival);
 - use fact that workload of M/G/1FCFS is same as M/G/1PS which can be run backwards in time in equilibrium; (PS means arrivals Share Processor.)
 - ▶ so $[M/G/1PS]^c$ can be used to provide Dominated CFTP.
- Connor & WSK (2015) also compare
 - © CFTP coupling when dominating process empties,
 - a faster CFTP coupling using upper and lower processes starting respectively at dominating process and at empty state.



Results (I)

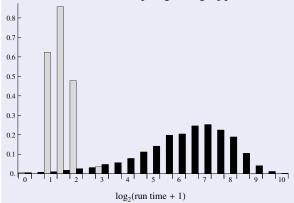
Histogram of customer numbers for M/M/c queue in equilibrium: arrival rate 10, service rate 2, and 10 servers, comparing theory (available for M/M/c queue) with results of Connor & WSK (2015) algorithm.



Results (II)

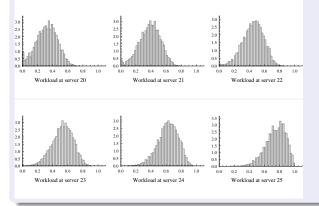
Comparison of log-run times for

- OFTP coupling when dominating process empties (solid bars),
- a faster CFTP coupling using upper and lower processes (grey bars).



Results (III)

Workload distributions of six most heavily loaded servers in an M/G/c=25 queue with Uniform(0,1) service time distribution and arrival rate 25.



Perfect Epidemics

- Even the simple case of deterministic S-I-R permits only *partial* closed-form solution;
- Suppose **only** the "removal times" are observed? Can the infection times be simulated perfectly? (Given basic parameters.)
- YES: work in progress by Connor and Kendall. Here is a GIF illustrating this for a real-life small-pox epidemic;
- (be clear about assumptions!)
- (indicate how the perfect simulation algorithm can be used as a high-dimensional integration device to enable simulation-based Bayesian inference!).

Conclusion

- You don't always have to put up with burn-in issues when doing MCMC;
- CFTP works even for significantly complex and relevant models of real-life phenomena;
- Of course really detailed models are still going to resist perfect simulation: but it will always be helpful to compare with a simpler model (using fewer parameters!).
- CFTP is clearly an important tool to be considered by the investigator seeking to do accurate and informative MCMC.

References I

- Brown, H. & May, M. (2004) Edward Teller in the Public Arena. *Physics Today*, **57**, 51–53.
- Connor, S.B. & WSK (2007) Perfect simulation for a class of positive recurrent Markov chains. Annals of Applied Probability, 17, 781–808.
- Connor, S.B. & WSK (2015) Perfect simulation of M/G/c queues. Advances in Applied Probability, 47, 1039–1063.
- Foss, S.G. & Tweedie, R.L. (1998) Perfect simulation and backward coupling. Stochastic Models, 14, 187–203.
- Huber, M.L. (2015) Perfect Simulation. Boca Raton: Chapman; Hall/CRC.
- Kiefer, J. & Wolfowitz, J. (1955) On the Theory of Queues With Many Servers. Transactions of the American Mathematical Society, 78, 1.
- Metropolis, N., Rosenbluth, A.W., Rosenbluth, M.N., Teller, A.H., & Teller, E. (1953) Equation of State Calculations by Fast Computing Machines. *The Journal of Chemical Physics*, **21**, 1087.
- Propp, J.G. & Wilson, D.B. (1996) Exact sampling with coupled Markov chains and applications to statistical mechanics. *Random Structures and Algorithms*, 9, 223–252.
- R Development Core Team (2010) R: A Language and Environment for Statistical Computing.
- Sigman, K. (2011) Exact simulation of the stationary distribution of the FIFO M/G/c queue. *Journal of Applied Probability*, **48**, 209–213.



References II

- WSK (1998) Perfect Simulation for the Area-Interaction Point Process. *Probability towards 2000* (Accardi, L. & Heyde, C.C. eds). Springer-Verlag, pp. 218–234.
- WSK (2004) Geometric ergodicity and perfect simulation. *Electronic Communications in Probability*, **9**, 140–151.
- WSK (2005) Notes on Perfect Simulation. Singapore: World Scientific, pp. 93-146.
- WSK (2015) Introduction to CFTP using R. Stochastic geometry, spatial statistics and random fields, Lecture notes in mathematics. Springer, pp. 405–439.
- WSK & Møller, J. (2000) Perfect simulation using dominating processes on ordered spaces, with application to locally stable point processes. Advances in Applied Probability, 32, 844–865.
- WSK & Thönnes, E. (1999) Perfect simulation in stochastic geometry. *Pattern Recognition*, **32**, 1569–1586.
- Zanella, G. (2015b) Bayesian Complementary Clustering, MCMC, and Anglo-Saxon Placenames (PhD Thesis).
- Zanella, G. (2015a) Random partition models and complementary clustering of Anglo-Saxon place-names. *Annals of Applied Statistics*, **9**, 1792–1822.



Image attributions

Image	Attribution	
Aristotle	After Lysippos	Public domain via Wikimedia Commons
Edward Teller	Lawrence Livermore National Laboratory restored by w:User:Greg L, Papa Lima Whiskey	CC BY-SA 3.0 via Wikimedia Commons
Perfect Ising	Result of code written by WSK	
Dead leaves	Result of code written by WSK	
Queues	https://covidposters.github.io/	Open source
Number of customers in $M/M/c$ queue	Result of code written by Stephen Connor	
Run-times for $M/M/c$ queues	Result of code written by Stephen Connor	
Epidemic	Result of code written by WSK	

These notes were produced from Thessaloniki-2024.qmd:

git repository information.			
Version: 0.0.36 (Tue May 7 16:11:07 2024 +0100)			
Author: Wilfrid Kendall < W.S. Kendall @warwick.ac.uk >			
Date: Tue May 7 17:55:14 2024 +0100			
Still this early version.			

