

Lecture 22

MH in practice

HW6 : now due 5/28

Lecture Notes VIII – Markov Chain Monte Carlo

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Notation ✓

Gibbs sampling ✓

The detailed balance ✓

Metropolis-Hastings sampling ←

Metropolis-Hastings algorithm

$$S(x, x') = \Pr(x' | x)$$

In Proposal distribution $S(x, x') \propto$ transition probability $x \rightarrow x'$
no need to be normalized, no need to be symmetric

How to choose S ?

1. Initialize x^0 with some arbitrary values
2. For $t = 1, 2, \dots$ we will sample sequentially $x^t | x^{t-1}$ as follows

2.1 Sample $x' \sim S(x^{t-1}, x')$

2.2 Compute acceptance probability

$$a(x^{t-1}, x') = \min \left(1, \frac{P(x')S(x', x^{t-1})}{P(x^{t-1})S(x^{t-1}, x')} \right) = \begin{cases} 1 & \text{if } \dots > \dots \\ \dots & \text{otherwise} \end{cases} \quad (10)$$

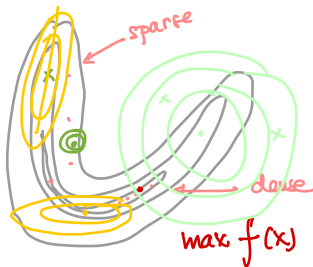
2.3 $x^t = \begin{cases} x' & \text{w.p. } a \\ x^{t-1} & \text{w.p. } 1 - a \end{cases}$

2.4 Every T steps (where T is a LARGE number), output x^t

↓
for n samples need nT MH steps



Hamiltonian MC



$S(x^t, x')$ - centered around x^t

- large variance $\Rightarrow \text{Prob}[\text{accept}] \ll 1$
 "spread"

- small variance $\Rightarrow$ mixing slow

$\rightarrow$ optimum by trial and error.

How to choose "mixing time" T !

Auto-covariance function

$x^1, x^2, \dots, x^t, \dots \sim \mathcal{P}$ desired

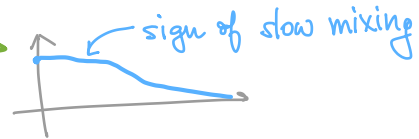
$$\hat{\Gamma}_s = \text{cov}[x^t, x^{t+s}]$$

$$\hat{\Gamma}_2 = \text{Var } X \text{ from } x^{1:N}$$

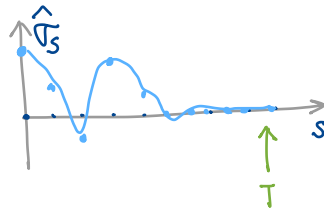
$$\hat{\mu} = E[x^t] = 0 \text{ w.l.o.g.}$$

$$\hat{\Gamma}_s = \frac{1}{N-s} \sum_{t=1}^{N-s} x^t x^{t+s}$$

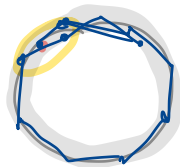
$x \in \mathbb{R}$



$$\hat{\Gamma}_2 = \hat{\Gamma}_0$$



$$\hat{\Gamma}_s \rightarrow 0 \text{ for large } s$$



Does it satisfy the detailed balance?

- ▶ If x' rejected ✓ $\Rightarrow x^t = x^{t-1}$
- ▶ If x' accepted $\Rightarrow x^t = x'$

$$P(x', x) = S(x', x) a(x', x) \quad (11)$$

$$P(x)P(x, x') = P(x)S(x, x') \min \left(1, \frac{P(x')S(x', x)}{P(x)S(x, x')} \right) \quad (12)$$

$$= \min (P(x')S(x', x), P(x)S(x, x')) \quad (13)$$

$$= P(x')P(x', x) \quad \text{by symmetry} \quad (14)$$

Recap: What we need to be able to do MH sampling

- ▶ To calculate $P(x)/P(x')$ but not P itself (okay not to have Z)
- ▶ To calculate $S(x, x')/S(x', x)$
- ▶ To **sample** from $S(x, x')$