

Day 12 in 2021s

SPPR Day 21

Markov Chain Monte Carlo

Metropolis Hastings Algorithm

Reading: Bishop PRML Sec 11.2.1

Markov chains

Sec 11.2.2

Metropolis
Hastings

- Topics:
- 1) MCMC with Propose-Accept transitions
 - 2) Random Walk proposals: intuition
 - 3) Detailed Balance, proof that RW works!
 - 4) Metropolis + Metropolis-Hastings Algorithms

Goal: Sample from target distribution
with PDF $p^*(z)$ over real $z \in \Omega \subseteq \mathbb{R}$

Challenge: Do NOT know how to evaluate $p^*(z)$,
only known up to multiplicative constant,
 $\tilde{p}(z)$ is evaluable

$$p^*(z) = c \tilde{p}(z) \text{ is NOT}$$

because $c = \frac{1}{\int \tilde{p}(z) dz}$ is intractable
(cannot do the integral)

Key idea: ratio of PDFs of two possible z values
can be evaluated exactly

$$\frac{p^*(z^1)}{p^*(z^2)} = \frac{\cancel{c} \tilde{p}(z^1)}{\cancel{c} \tilde{p}(z^2)} = \frac{\tilde{p}(z^1)}{\tilde{p}(z^2)} \quad \text{easy to compute!}$$

Want to design Markov proposal that
some how uses ratios

Markov Transition Design: Propose then ^{accept/}reject

Transition distribution \mathcal{Z} takes current "state" value z_t , and produces new "state" value z_{t+1}

Idea: Propose new value $z' \sim Q(z'|z_t)$ then decide to accept w. prob $A(z', z_t)$

$$z_{t+1} = \begin{cases} z' & \text{if } u < A(z', z_t) \\ z_t & \text{otherwise} \end{cases}$$

where $z' \sim Q(z'|z_t)$
 $u \sim \text{Unif}([0, 1])$

Example

Start: $z_1 = 0.3$

\mathcal{Z}

- 1) Sample $z' \sim Q(\cdot|z_1)$. $z' = -1.8$
- 2) Sample $u \sim \text{Unif}([0, 1])$. $u = 0.3$
- 3) Eval $A(z', z_1) = 0.8$. $u < A$ so ACCEPT

$z_2 = -1.8$

\mathcal{Z}

- 1) Sample $z' \sim Q(\cdot|z_2)$. $z' = 2.7$
- 2) Sample $u \sim \text{Unif}([0, 1])$. $u = 0.91$
- 3) Eval $A(z', z_2) = 0.01$. $u > A$ so REJECT

$z_3 = -1.8$

(and keep going to get z_1, z_2, \dots, z_S)

Remember, A is an accept probability threshold

We know $0 \leq A \leq +\infty$

$A = 0.0$ means no chance of accept
ALWAYS REJECT

$A \in (0, 0.5)$ means can accept,
but likely reject

$A \in (0.5, 1)$ means can reject
but likely accept

$A \geq 1$ means ALWAYS ACCEPT

So sometimes we write

$$P(\text{accept} | z', z_t) = \min(A(z', z_t), 1)$$

any value ≥ 1
is just set to 1

Proposal Idea: Random Walk

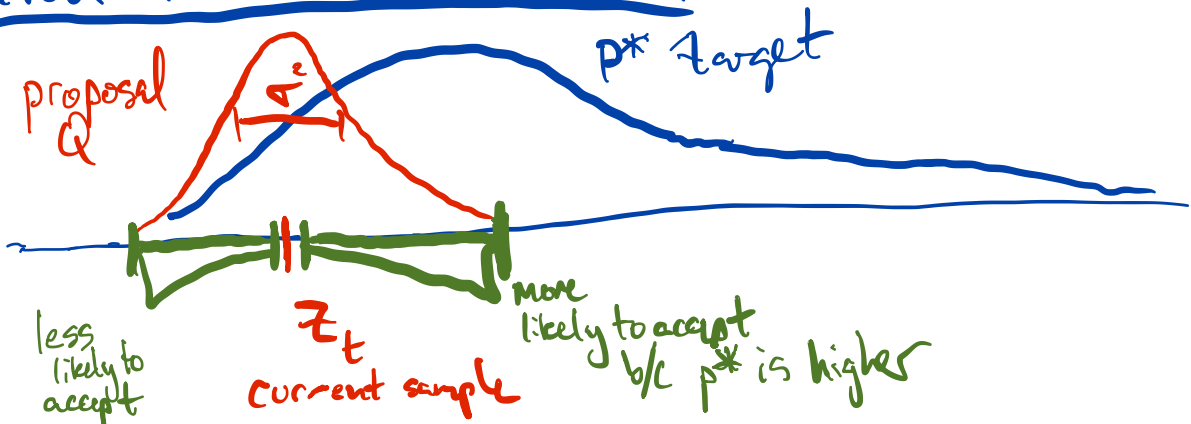
Random Walk

$$Q(z' | z_t) = \text{Normal}\left(z' \mid \bar{z}_t, \frac{\text{var}}{\sigma^2}\right)$$

you pick $\sigma^2 > 0$ hyperparam.

why: easy to sample
likely to propose "nearby"

Intuition for Random Walk



let's try

$$A(z', z_t) = \frac{\tilde{p}(z')}{\tilde{p}(z_t)} = \frac{p^*(z')}{p^*(z_t)}$$

remember ratios of p^* can be computed exact

given current z_t , and proposed value z' ,

$p^*(z')$ — than $p^*(z_t)$	ratio $\frac{p^*(z')}{p^*(z_t)}$	proposal will be
Much smaller	$\ll 0.5$	likely rejected
a bit smaller	in $(0.5, 1)$	likely accepted
Same or larger	≥ 1	certainly accepted

Intuitively, random walk proposal Q plus $\frac{p^*}{p}$ ratio A makes sense, but how to show this will create Markov chain w/ stationary distr. p^* ?

Recall, p^* is stationary for transition \mathcal{T} if:

$$\text{for all } z_{t+1} \in \Omega: p^*(z_{t+1}) = \int_{z_t} p^*(z_t) \mathcal{T}(z_{t+1} | z_t) dz_t$$

Can we show that when \mathcal{T} uses rand walk Q and accept proba A , that this is satisfied?

First, define PDF. $\mathcal{T}(z_{t+1} | z_t) = \min\left(1, \frac{p^*(z_{t+1})}{p^*(z_t)}\right) Q(z_{t+1} | z_t)$ when $z_{t+1} \neq z_t$

Next, show for all $z_a \neq z_b$

$$p^*(z_a) \mathcal{T}(z_b | z_a) = p^*(z_b) \mathcal{T}(z_a | z_b)$$

$$\cancel{\min\left(\frac{p^*(z_a)}{p^*(z_b)}, \frac{p^*(z_b)}{p^*(z_a)}\right) Q(z_b | z_a)} = \cancel{\min\left(\frac{p^*(z_b)}{p^*(z_a)}, \frac{p^*(z_a)}{p^*(z_b)}\right) Q(z_a | z_b)}$$

and we know $Q(z_b | z_a) = Q(z_a | z_b)$ because $\mathcal{N}(z_a | z_b, \Sigma) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{|\Sigma|}} e^{-\frac{1}{2} z^T \Sigma^{-1} z}$ Symmetric $= \mathcal{N}(z_b | z_a, \Sigma)$

Remember, for any $a > 0, b > 0$
 $a \cdot \min(1, \frac{b}{a})$

$$= a \cdot 1 \text{ if } b > a = \min(a, b)$$
$$a \cdot \frac{b}{a} \text{ if } b \leq a$$

Now, for

$$\cancel{P^*(z_t)} \tilde{\Gamma}(z_{t+1} | z_t) = \cancel{P^*(z_{t+1})} \tilde{\Gamma}(z_t | z_{t+1})$$

$$z_t = z_{t+1} \\ = \tilde{z}$$

$$z_t = z_{t+1}$$

$$z_t = z_{t+1}$$

$$\tilde{\Gamma}(\tilde{z} | \tilde{z}) = \tilde{\Gamma}(\tilde{z} | \tilde{z})$$

true
by symmetry

So we have

shown for all $z_a \in \Omega, z_b \in \Omega$, that random walk transition $\tilde{\Gamma}$ satisfies:

$$P^*(z_a) \tilde{\Gamma}(z_b | z_a) = P^*(z_b) \tilde{\Gamma}(z_a | z_b)$$

We call this the DETAILED BALANCE condition

$$z_a \sim P^* \quad \text{has same joint probab as} \quad z_b \sim P^*$$
$$z_b \sim \tilde{\Gamma}(\cdot | z_a) \quad z_a \sim \tilde{\Gamma}(\cdot | z_b)$$

but, how does this help show P^* is stationary of $\tilde{\Gamma}$?

well, Proving stationary if τ satisfies DETAILED BALANCE

$$p^*(z_{t+1}) = \int p^*(z_t) \tau(z_{t+1}/z_t) dz_t$$

$$= \int p^*(z_{t+1}) \tau(z_t/z_{t+1}) dz_t$$

by detailed balance

$$= p^*(z_{t+1}) \int \tau(z_t/z_{t+1}) dz_t$$

bringing out term const
wrt z_t

$$= p^*(z_{t+1})$$

because τ is a PDF
and must integrate
to 1

Thus, p^* is stationary distribution,
because it meets required definition.

METROPOLIS MCMC Algorithm

Initialize $z_1 \in \Omega$

for t in $1, 2, \dots, S-1$

1) $z' \sim Q(\cdot | z_t)$

2) $u \sim \text{Unit}([0, 1])$

3) $z_{t+1} = \begin{cases} z' & \text{if } u \leq \frac{\tilde{p}(z')}{\tilde{p}(z_t)} \\ z_t & \text{otherwise} \end{cases}$

return $[z_1, z_2, \dots, z_S]$

Assumes that Q valid PDF over Ω
easy to sample
easy to evaluate PDF
must be symmetric

$$Q(z_a | z_b) = Q(z_b | z_a)$$

for all $z_a \in \Omega, z_b \in \Omega$

$$\frac{\tilde{p}(z')}{\tilde{p}(z_t)} = A(z', z_t)$$

if S large enough,
can consider
 z_{S-B}, \dots, z_S as samples of p^*

METROPOLIS-HASTINGS MCMC Algorithm

Initialize $z_1 \in \Omega$

for t in $1, 2, \dots, S-1$

1) $z' \sim Q(\cdot | z_t)$

2) $u \sim \text{Unit}([0, 1])$

3) $z_{t+1} = \begin{cases} z' & \text{if } u < \frac{\tilde{p}(z')}{\tilde{p}(z_t)} \frac{Q(z_t | z')}{Q(z' | z_t)} \\ z_t & \text{otherwise} \end{cases}$

return $[z_1, z_2, \dots, z_S]$

Assumes only that
 Q is valid PDF over Ω
easy to sample
& easy to evaluate PDF

$$\frac{\tilde{p}(z')}{\tilde{p}(z_t)} \frac{Q(z_t | z')}{Q(z' | z_t)}$$

can show this A also
satisfies DETAILED BALANCE

Sanity check: What if we use $Q = p^*$ in Met-Hodgins

$$\text{Accept ratio } A = \frac{\tilde{p}(z')}{\tilde{p}(z_t)} \frac{p^*(z_t)}{p^*(z')} = \frac{\tilde{p}(z') \tilde{p}(z_t) c}{\tilde{p}(z_t) \tilde{p}(z') c} = 1$$

So we'd always accept! Makes sense.