

Welcome

# COMP 136

## Day 21: MCMC

CP4 - implement!  
forward alg HMMs  
viterbi alg HMM

~5 hours  
~~20 hr~~

with  
Metropolis's  
Hastings

### Agenda:

- 1 - quiz3 recap
- 2 - Q&A

### Upcoming Due Dates

- HW4 released! (due Wed at 11:59pm ET)
- CP4 released, due next Tues (that is late deadline)
- = Q4 late next week

= HWS, QS, CPS soon

details about final

# Random Walk Samples (uses Metropolis's MCMC algorithm)

$r.v. z \in \mathbb{R}^D$   
real valued

$z_1$  initial

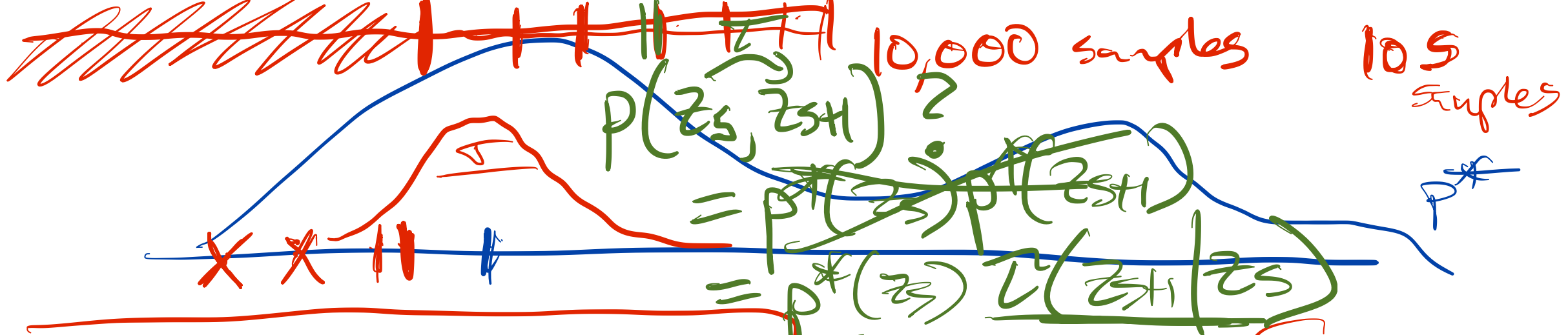
$$z' \sim Q(z' | z_t) = \mathcal{N} \left( \begin{matrix} \text{mean} \\ z_t, \end{matrix} \begin{matrix} \text{var} \\ \sigma^2 \end{matrix} \right)$$

0.1  
↓

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

return  $z_1, z_2, \dots, z_s$

st  $z_s \sim p^*$



$z_1, z_2, z_3, z_4, \dots, z_s, z_{s+1}, \dots, z_{2s}, \dots$

$z_t$  not indep. of  $z_{t-1}$

"thinning"  
keep every 10<sup>th</sup>  
100<sup>th</sup>

with enough samples

$\exists S \gg 1$ , st.  $p(z_s) = p^*(z)$  if  $S$  is large enough

"discard burn-in"  
keep after  $B$ <sup>th</sup>

Recall:

cannot eval  $p^*(z)$

can eval  $\tilde{p}(z)$

$$p^*(z) = c \tilde{p}(z)$$

valid PDF      scalar pos      scalar pos  
 (under  $p^*(z)$ )      (under  $c$ )      (under  $\tilde{p}(z)$ )

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

accept  $\tilde{p}(z) \gg p(z)$   
 accept  $0.5 < 1$   $\tilde{p}(z) < \tilde{p}(z')$   
 most likely reject  $\tilde{p}(z) < \tilde{p}(z')$

Usually: what we want target to be  
is posterior of some r.v. given data

if  $\mathcal{Z}$  is ergodic  
 $\exists$  one unique  $\mathbb{P}^*$   
stationary  $\mathbb{P}^*$

$z$ : param of model

Prior:  $p(z)$

evaluable!

Lik:  $p(x/z) = \prod_{n=1}^N \mathcal{N}(x_n/z, \beta^{-1})$

evaluable!

Posterior:  $p(z/x) = \frac{1}{p(x)} p(z) p(x/z)$

$\mathbb{P}^*(z)$

number

Prior:  $z \sim \text{GMM}(\pi, \mu, \sigma)$

Lik:  $x_n | z \sim \mathcal{N}(\overset{\text{mean}}{z}, \overset{\text{prec}}{\beta^{-1}})$

Given  $x_1, \dots, x_n$ , sample  $p(z | x_1, \dots, x_n)$

Random Walk accept proba:

$$A: \frac{\tilde{p}(z')}{\tilde{p}(z_t)} = \frac{\text{GMM}(z' | \pi, \mu, \sigma) \prod_n \mathcal{N}(x_n | z', \beta^{-1})}{\text{GMM}(z_t | \pi, \mu, \sigma) \prod_n \mathcal{N}(x_n | z_t, \beta^{-1})}$$

2 things you can know:

- eval pdf function ( $\hat{P}$ ) up to c

- sample from a distribution

Normal PDF

$$\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

Exponential PDF ✓

sample?

~~np and n~~

$u \sim \text{Unif}(0, 1)$

CDF not analytical invertible

# Metropolis

RW

$$N(z_t | -)$$

Q valid PDF  
easy sample

symmetry  $Q(z_t | z_b) = Q(z_b | z_t)$   
for all  $z_t, z_b$

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


more general

# Metropolis Hastings

Q valid PDF  
easy sample  
easy eval PDF

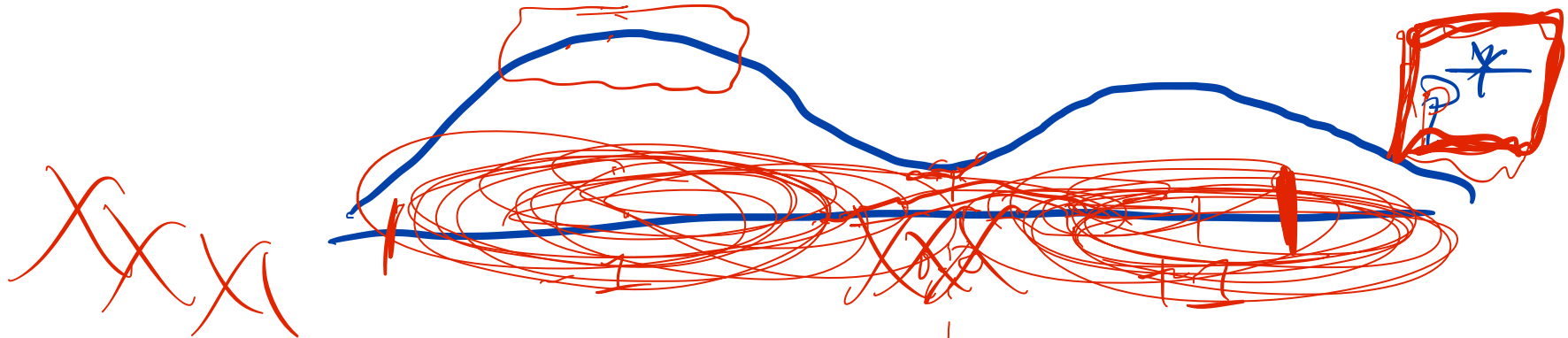
NO SYM REQ'D

$$\tilde{p}(a) Q(b|a) = \tilde{p}(b) Q(a|b)$$

special case  
Q from Hamiltonian MC

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

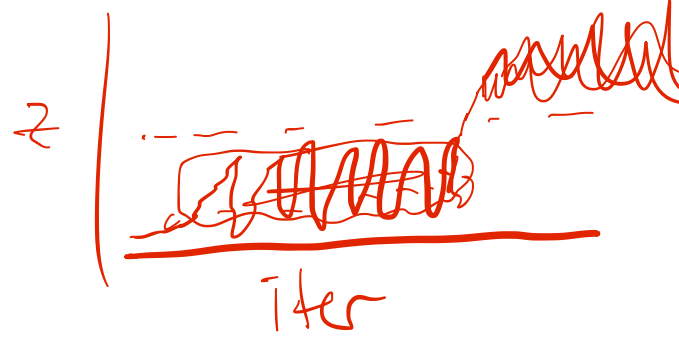
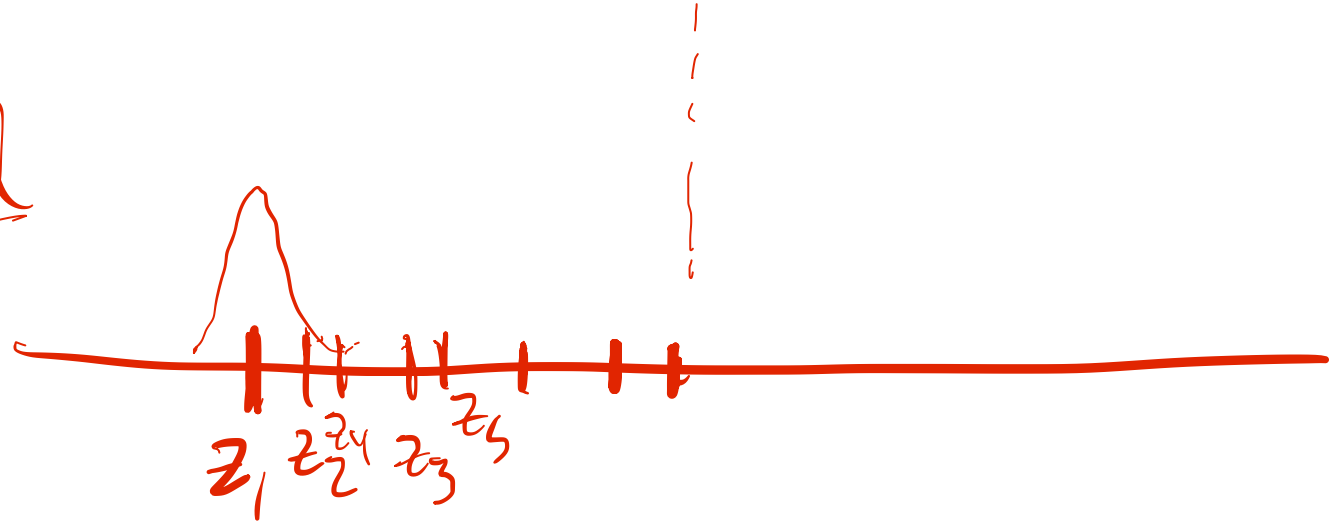




Question: in future walk thru of RW?

~~X X X~~

RW with  $\Delta$  small



RW  $\Delta$  large

