# Introduction to Monte Carlo Markov Chains (Part 2)

Class 20

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## But first ... scripts vs programs

```
def myfunc(x,y,z,...):
    do a bunch of stuff here
    return f(x,y,z,...)
```

```
Program, Function or Class memory space

x
y
z
.
Intermediate calcs
f(x,y,z...)
```

```
Script
memory space

X

Y

Z

Intermediate calcs
f(x,y,z...)

ipython kernel
```

# My recommendations

Always write first drafts of code in script form.

Always do your code development in ipython.

This makes debugging straightforward

Then ... copy your useful (debugged) code into a function or class for long-term use.

• MCMC is a technique for taking a "biased random walk" through the space of  $\pi(\vec{a})$ 

 It is "random" in the sense that we will draw from probability distributions to guess which way to go.

• It is "biased" in the sense that we will allow  $\pi(\vec{a})$  to influence how often we choose the uphill direction.

## Metropolis - Hastings

 Last time we implemented a simple Metropolis-Hastings algorithm for our sampler.

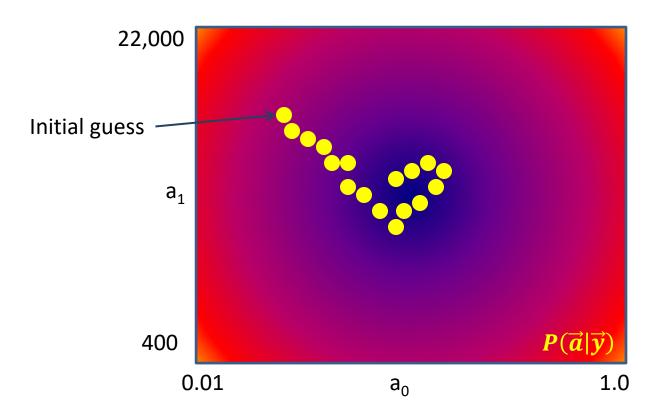
#### Key points:

- ergodic chains
- transition probability satisfies detailed balance

#### Challenges of M-H

- for M free parameters there are a minimum of M parameters in the transition probability to be tuned by the user.
- the autocorrelation time (# of samples required to achieve independent samples of the posterior distribution) is not optimally short.

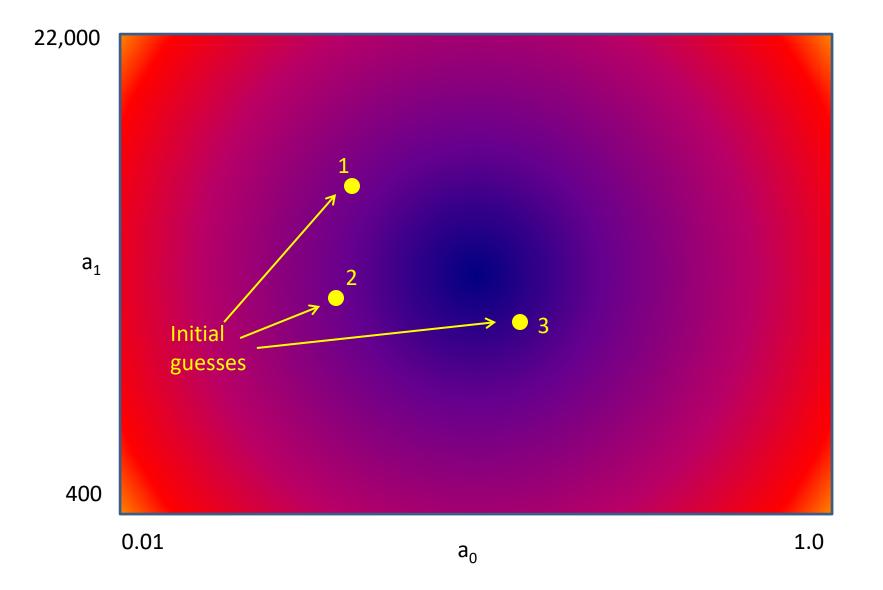
 M-H samplers require choice of step size for each parameter of interest

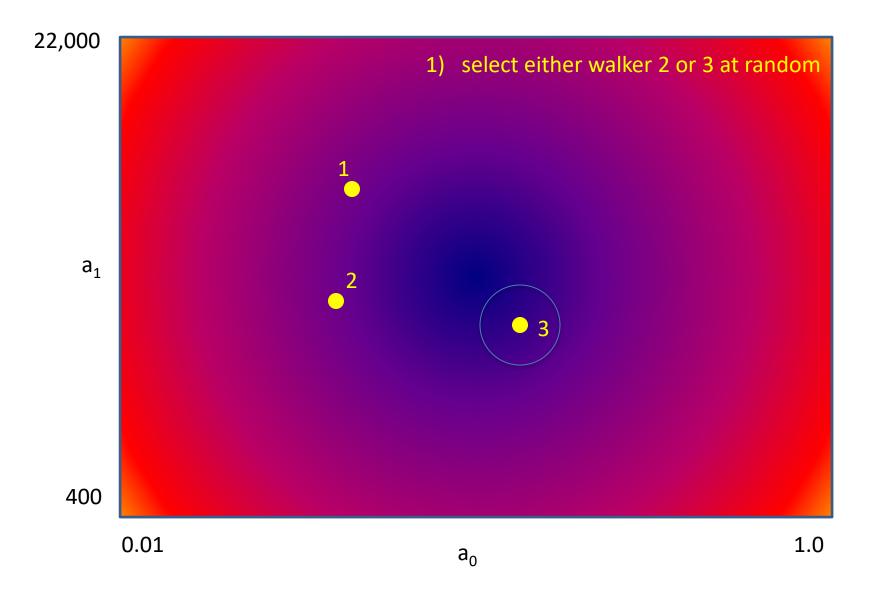


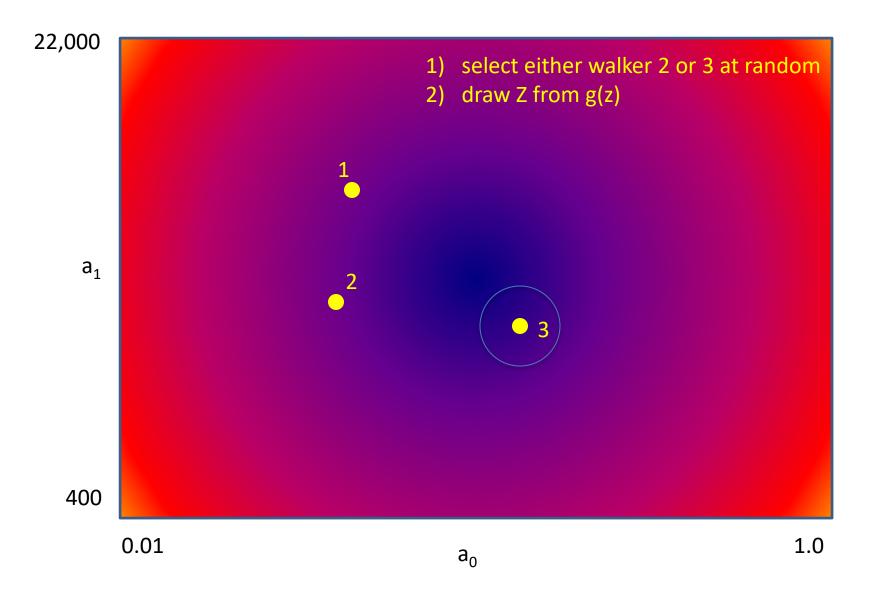
Over time, the chain will fill in the space with a density of visits proportional to the posterior probability distribution.

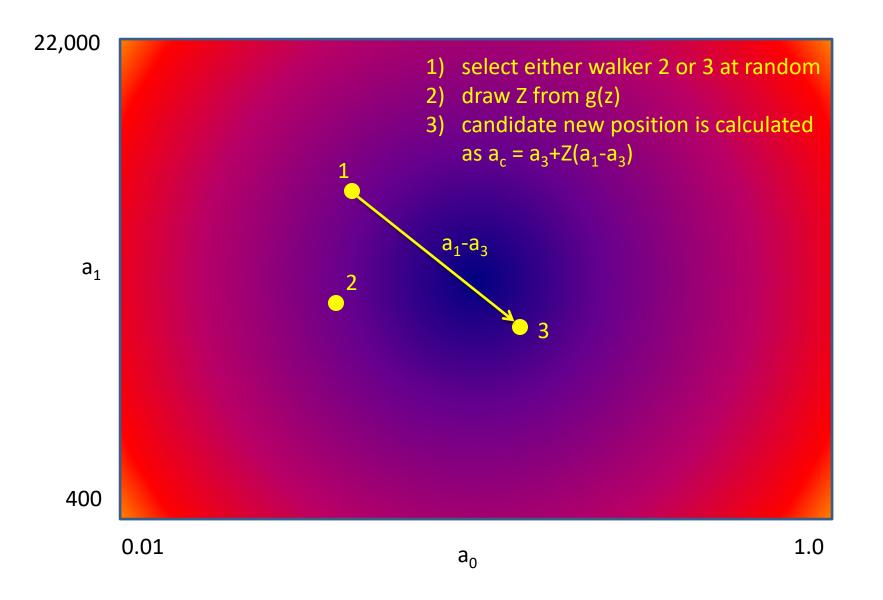
# The Stretch Move Approach

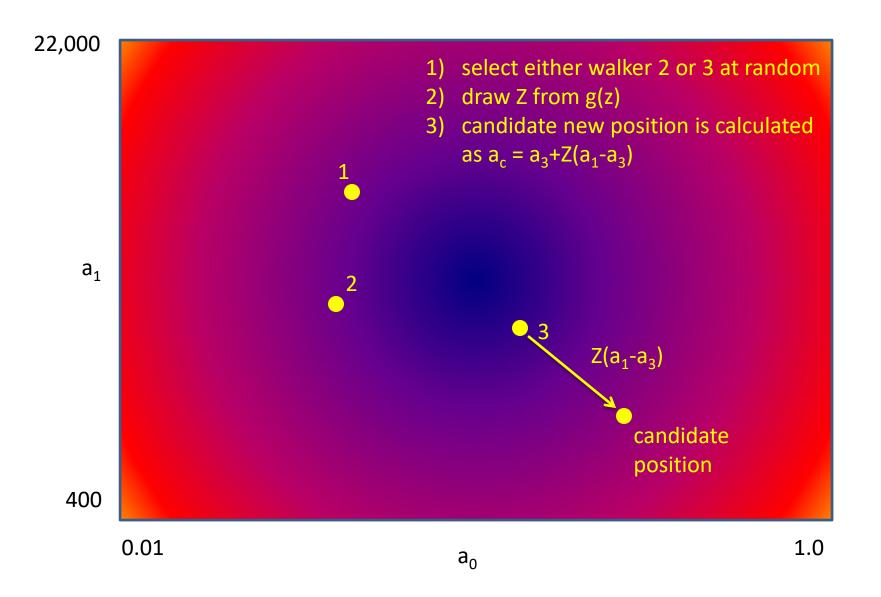
- an affine-invariant ensemble sampling algorithm
- still generates ergodic chains
  - heads of chains are called "walkers"
- still satisfies detailed balance
- only one free parameter to set
  - set it to 2

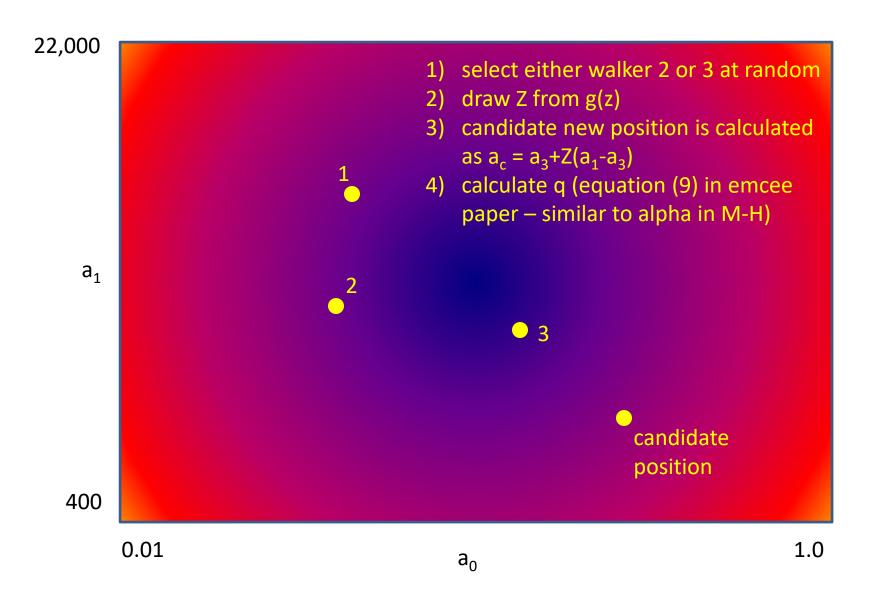


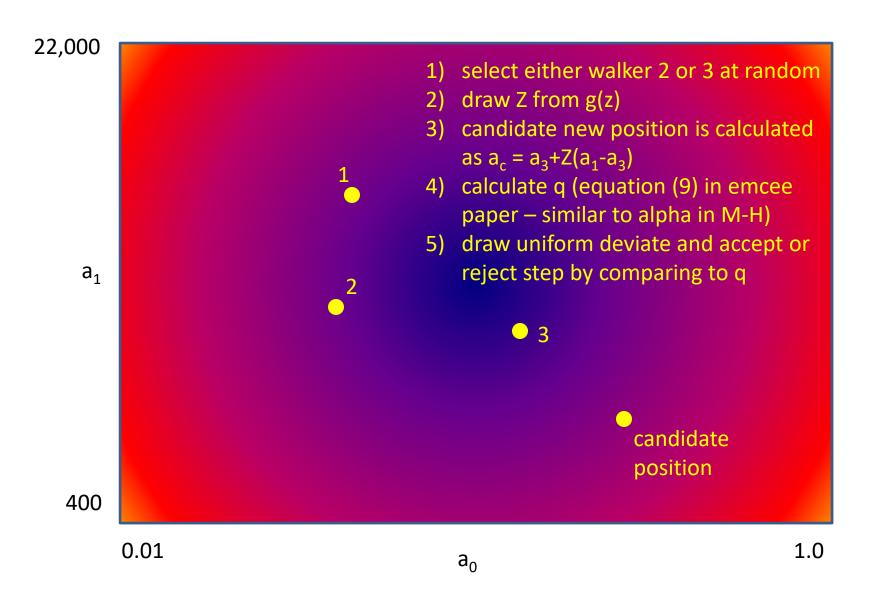


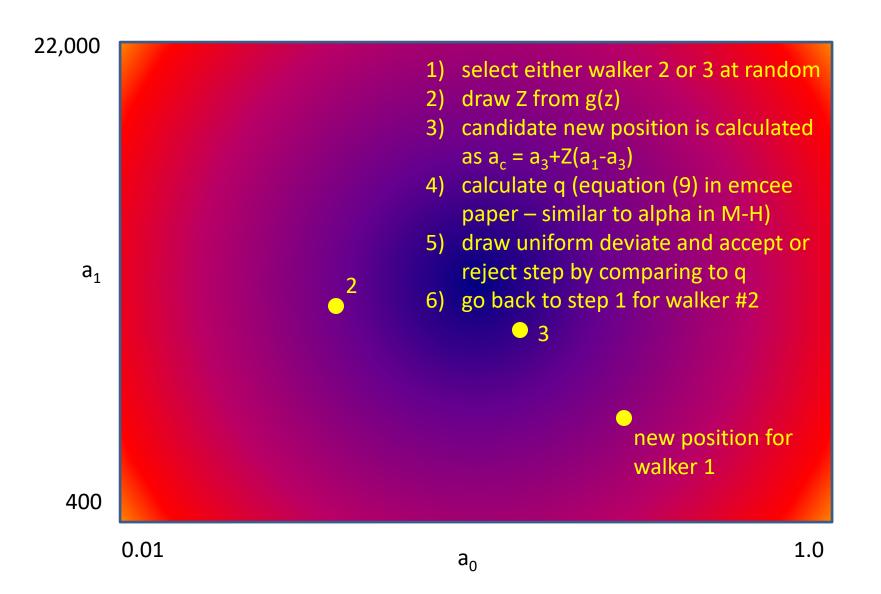












## **Important Points**

- For detailed balance to be preserved these steps must occur in series. Deal with one walker at a time.
- Recipe is given in paper for parallelizing process.
- If things are going well, acceptance rate will be between 0.2 and 0.5.
- The technique works best with many walkers at minimum 2x the number of free parameters (but go with something like 10x if possible)
- try downloading the emcee package with pip
  - pip install emcee
  - see http://dan.iel.fm/emcee/current/

### Exercise #1

 Re-work the two MCMC problems from last class using the emcee package.

Do it again using the emcee interface to Imfit