

Title: Computational Physics - Lecture 11

Date: Oct 17, 2018 01:00 PM

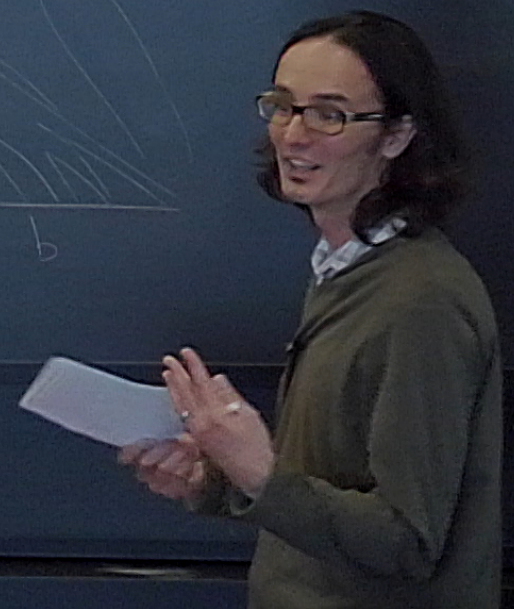
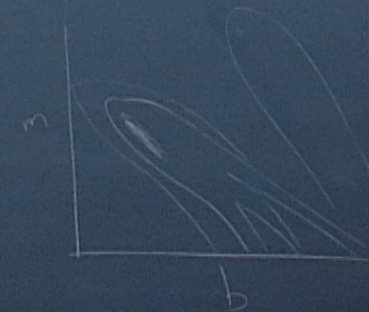
URL: <http://pirsa.org/18100052>

Abstract:

Last time

- foreground-background

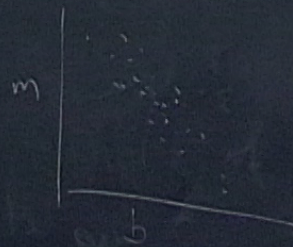
$$p(y | m, b, P_{\text{bad}}, \gamma, V) = P_{\text{bad}} \cdot \frac{1}{\sqrt{2\pi(V\sigma^2)}} \exp\left(-\frac{(y_i - \gamma)^2}{2(V\sigma^2)}\right) \\ + (1 - P_{\text{bad}}) \cdot \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y_i - mx - b)^2}{2\sigma^2}\right)$$



Markov Chain Monte Carlo (MCMC)

Solves

- have a probability dist. $f(\text{params})$
- want a sampling of params



$$\Theta = (b, m)$$

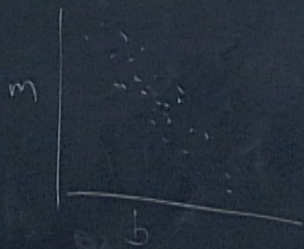
$$v(b, m, P_{\text{obs}}, Y, V)$$

Idea

Markov Chain Monte Carlo (MCMC)

Solves

- have a probability dist $f(\text{params})$
- want a sampling of params



$$\theta = (b, m)$$

or (b, m, ρ_{out}, Y, V)

Idea

- move a "particle"

$$\theta_{\text{new}} = \theta + \underbrace{N(0, \sigma_{\theta}^2)}_{\text{proposal distrib}}$$

- compute prob at θ_{new} $P_{\text{new}} = P(\theta_{\text{new}})$

- always keep θ_{new} that improve P_{new} vs $P(\theta)$

Sometimes keep θ_{new} with lower P_{new} : $\propto \frac{P(\theta_{\text{new}})}{P(\theta)}$

- in the long run, particle posns θ are fair samples

m

$$\theta = (b, m)$$

$$\text{or } (b, m, P_{\text{ord}}, Y, V)$$

- always keep θ_{new} that improve P_{new} vs $P(\theta)$
- sometimes keep θ_{new} with lower P_{new} : $\propto \frac{P(\theta_{\text{new}})}{P(\theta)}$
- in the long run, particle posns θ are fair samples

Metropolis-Hastings
1953 1970

```
chain = []
params = [50, 2]
jumpsizes = [10, 0.1]
prob = prob-function(params)
```

```
function line-prob(b, m)
  exp(-line-hd(b, m))
end
```

```
for L in 1:Nsteps
  params-new = params + randn(2) * jumpsizes
```

m

$$\Theta = (b, m)$$

$$\text{or } (b, m, P_{\text{ord}}, Y, V)$$

- always keep Θ_{new} that improve P_{new} vs $P(\Theta)$
- sometimes keep Θ_{new} with lower P_{new} : $\propto \frac{P(\Theta_{\text{new}})}{P(\Theta)}$
- in the long run, particle positions Θ are fair samples

Metropolis-Hastings
1953 1970

```

chain = []
params = [50, 2]
jumpsizes = [10, 0.1]
prob = prob-function(params)
for l in 1:Nsteps
  params_new = params .+ randn(2) .* jumpsizes
  prob_new = prob-function(params_new)

```

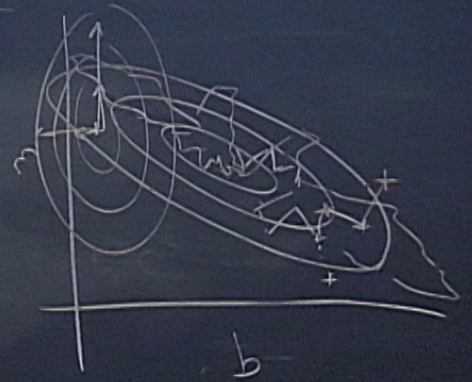
```

function line-prob(b, m)
  exp(-line-hl(b, m))
end

```

→ params

```
if (prob_new / prob  $\geq$  rand(Float64))
  #accept
  params = params_new
  prob = prob_new
end
append!(chain, params)
end
return chain
```



```

if (prob_new / prob >= rand(Float64))
    #accept
    params = params_new
    prob = prob_new
end
append!(chain, params)
end
return chain

```

