

$$p(a) = \int p(a, b) db$$

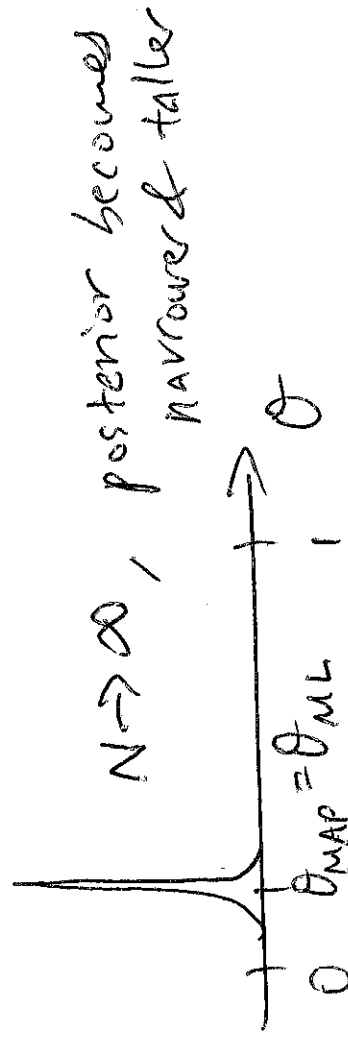
$$p(a|c) = \int \underbrace{p(a, b|c)} db$$

$$p(a, b) = p(a|b) p(b)$$

$$p(a, b|c) = \underbrace{p(a|b, c) p(b|c)}_{\text{always true}}$$

$$p(a, b|c) \neq p(a|c) p(b|c)$$

if  $a$  &  $b$  are  
conditionally  
independent  
given  $c$



$$p(y|x, \theta) = \frac{p(x, y|\theta)}{\underbrace{p(x|\theta)}} \propto p(x, y|\theta)$$

$$\underset{\text{parameter}}{\operatorname{argmax}} \underbrace{p(\text{parameter} | \text{data})}_{\text{mode of distribution } p(z)} \leftarrow \text{MAP}$$

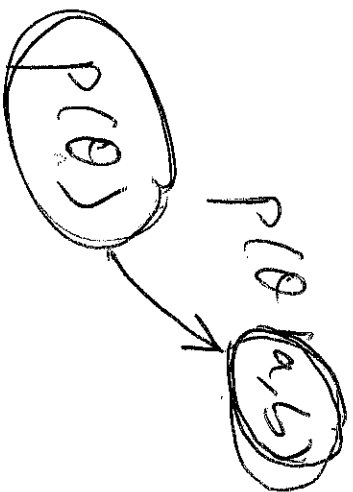
$$\underset{\text{parameters}}{\operatorname{argmax}} p(\text{data} | \text{parameters}) \leftarrow \text{ML}$$

$$\underset{z}{\operatorname{argmax}} p(z) = \text{mode of distribution } p(z)$$

$$\int p(\theta|D) d\theta = \int \frac{k(N_0, N_1, a, b) \text{Beta}(\theta|N_1+a, N_0+b)}{p(D)} d\theta$$

$$1 = \frac{k(N_0, N_1, a, b)}{p(D)}$$

$p(x) = \text{pdf}$  if  $x$  is continuous  
~~pdf~~  $p(x) = \text{pmf}$  if  $x$  is discrete



$$x \sim \mathcal{N}(0, 1)$$

$$p(x) > 0$$

$$\int p(x) dx = 1$$

$$\text{OR } p(x | \mu=0, \sigma^2=1)$$

Baseball

$$\operatorname{argmax}_{a,b} p(\text{data} | a, b)$$

Generative

$$X \sim N(0, 5)$$

We know  $p(X)$ , so we can sample from  $p(X)$

Feature  $X$ , want to predict  $Y$

We know  $p(X, Y)$ , so we can sample from  $p(X, Y)$

Beta-Binomial

$$p(D) = \int p(D, \theta) d\theta$$

$$= \int p(\theta) p(D | \theta) d\theta$$

$\uparrow$  Beta dist  
 $\uparrow$  Binomial likelihood

= function of  $a, b, N_0, N_1$

$$= p(D | a, b)$$

$$p(\text{John hits the next ball} | \text{history}) = \frac{20 + 70}{100 + 70 + 200} = A$$

$$p(\text{John does not hit the ball} | \text{history}) = 1 - A$$

$$= 1 - \frac{20 + 70}{100 + 70 + 200}$$

$$= \frac{80 + 200}{100 + 70 + 200}$$