

6.20  $\{x_i\}_{i=1}^N$

$$p(x|a,b) \sim \frac{d\phi}{dx} = \frac{b}{b^2 + (x-a)^2}$$

We drop the conditioning of  $x \in [-1,1]$  to avoid cluttering the notation.

but since  $p(\cdot|a,b)$  is a probability density we know

$$p(x|a,b) = \frac{1}{\underbrace{\int_{-1}^1 p(x) dx}} \cdot \frac{b}{b^2 + (x-a)^2} \stackrel{?}{=} C(a,b)^{-1} \frac{b}{b^2 + (x-a)^2}$$

$$= -\arctan\left(\frac{a-x-1}{b}\right) + \arctan\left(\frac{a-x+1}{b}\right) = C(a,b)$$

$$p(\{x_i\}|a,b) = \prod_{i=1}^N p(x_i|a,b) = \prod_{i=1}^N C(a,b)^{-1} \frac{b}{b^2 + (x_i-a)^2}$$

From Bayes theorem we know

$$p(a,b|\{x_i\}) = \frac{p(\{x_i\}|a,b) p(a,b)}{p(\{x_i\})}$$

for our prior we let  $p(a,b)$  be uniformly distributed on  $-1 < a < 1$   $0 < b < \infty$ .

$p(\{x_i\})$  is irrelevant as it cancels out.

as long as  $(a',b')$  is in  $[-1,1] \times (0,\infty)$  then it's 0

$$\text{we consider } \frac{p(a',b'|\{x_i\})}{p(a,b|\{x_i\})} = \frac{p(\{x_i\}|a',b') p(a',b')}{p(\{x_i\}|a,b) p(a,b)} = \frac{p(\{x_i\}|a',b')}{p(\{x_i\}|a,b)}$$

$$= \frac{b'}{b} \frac{C(a,b)}{C(a',b')} \frac{b^2 + (x_i-a)^2}{b'^2 + (x_i-a')^2}$$