

## Mixture of priors

Summary: Given  $f(\mu) = \alpha g(\mu) + \beta h(\mu)$  [mixed prior]

Find: -  $g(\mu|x)$  [posterior due to  $g(\mu)$ ]

-  $h(\mu|x)$  [posterior due to  $h(\mu)$ ]

Then  $f(\mu|x) = \alpha(x) g(\mu|x) + \beta(x) h(\mu|x)$

where  $\alpha(x)$  prop. to  $\alpha \int f(x|\mu) g(\mu) d\mu = \alpha g(x)$

$\beta(x)$  prop. to  $\beta \int f(x|\mu) h(\mu) d\mu = \beta h(x)$

$\alpha(x) + \beta(x) = 1$

Example: Assume  $x|\mu \sim N(\mu, \sigma^2)$

and  $g(\mu) : N(\beta, \tau^2)$

$h(\mu) \propto 1$  (improper prior)

We know:

$$g(\mu|x) : N\left(\frac{\sigma^2\beta + \tau^2x}{\sigma^2 + \tau^2}, \frac{\sigma^2\tau^2}{\sigma^2 + \tau^2}\right)$$

$$h(\mu|x) : N(x, \sigma^2) \quad [\text{as if } \tau^2 \rightarrow \infty]$$

$$\beta(x) \propto \beta \int_{-\infty}^{+\infty} f(x|\mu) \cdot 1 \cdot d\mu = \beta$$

$$q(x) \propto \alpha \int_{-\infty}^{+\infty} f(x|\mu) g(\mu) d\mu = \alpha \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi}\tau} e^{-\frac{(\mu-\beta)^2}{2\tau^2}} d\mu$$

Rearrange to make form  $\star e^{-\frac{[\mu-\star]^2}{2\star^2}}$

Consider  $\frac{(x-\mu)^2}{2\sigma^2} + \frac{(\mu-\beta)^2}{2\tau^2}$

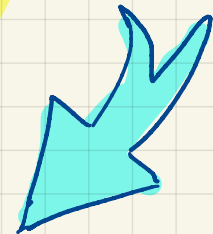
$$= \frac{x^2 + \mu^2 - 2\mu x}{2\sigma^2} + \frac{\mu^2 + \beta^2 - 2\mu\beta}{2\tau^2}$$

$$= \frac{\sigma^2 x^2 + \tau^2 \mu^2 - 2\mu \sigma^2 x + \sigma^2 \mu^2 + \tau^2 \beta^2 - 2\mu \tau^2 \beta}{2\sigma^2 \tau^2}$$

$$= \frac{\mu^2 [\sigma^2 + \tau^2] - 2\mu [\sigma^2 x + \tau^2 \beta] + \sigma^2 x^2 + \tau^2 \beta^2}{2\sigma^2 \tau^2}$$

$$= \frac{\left[ \mu - \frac{\sigma^2 x + \tau^2 \beta}{\sigma^2 + \tau^2} \right]^2}{\frac{2\sigma^2 \tau^2}{\sigma^2 + \tau^2}} - \frac{\left( \frac{\sigma^2 x + \tau^2 \beta}{\sigma^2 + \tau^2} \right)^2}{\frac{2\sigma^2 \tau^2}{\sigma^2 + \tau^2}} + \frac{\sigma^2 x^2 + \tau^2 \beta^2}{2\sigma^2 \tau^2}$$

$$= \frac{[\mu - \text{cloud}]^2}{\frac{2\sigma^2 \tau^2}{\sigma^2 + \tau^2}} + \frac{(x-\beta)^2}{2(\sigma^2 + \tau^2)}$$



$$\begin{aligned}
 \alpha(x) &\propto \alpha \int \frac{\sqrt{\sigma^2 + z^2}}{\sqrt{2\pi} \sigma z \sqrt{\sigma^2 + z^2}} e^{-\frac{[\mu - \text{cloud}]^2}{2\sigma^2 z^2 / (\sigma^2 + z^2)}} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x - \beta)^2}{2(\sigma^2 + z^2)}} d\mu \\
 &= \underbrace{\alpha \frac{1}{\sqrt{2\pi} \sqrt{\sigma^2 + z^2}} e^{-\frac{(x - \beta)^2}{2(\sigma^2 + z^2)}}}_{N(\beta, \sigma^2 + z^2)} [f(x) \text{ due to } g(\mu)]
 \end{aligned}$$

$$\alpha(x) = \frac{\alpha N(\beta, \sigma^2 + z^2)}{\alpha N(\beta, \sigma^2 + z^2) + \beta} \qquad \beta(x) = \frac{\beta}{\alpha N(\beta, \sigma^2 + z^2) + \beta}$$

We learned something:

$$f(x|\mu) \sim N(\mu, \sigma^2)$$

$$\mu \sim N(\beta, \tau^2)$$

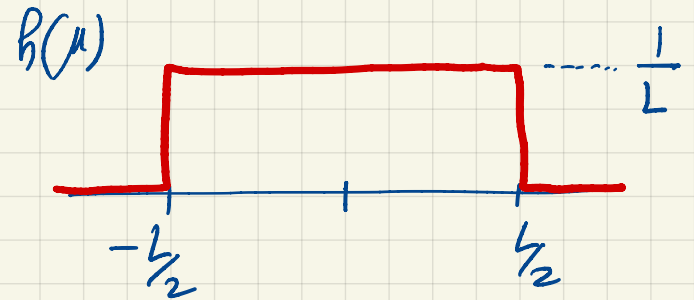
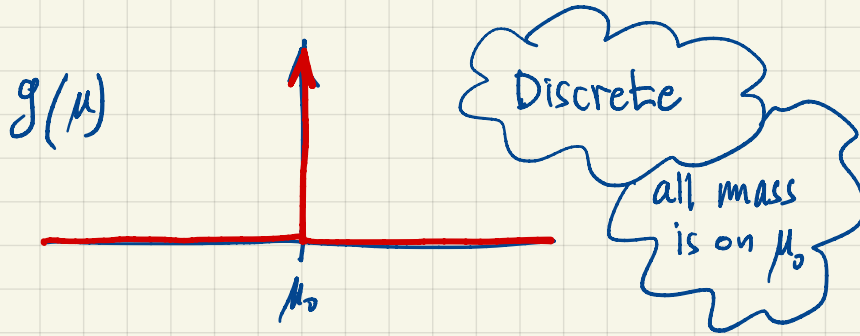
$$f(x) = \int_{-\infty}^{+\infty} f(x|\mu) f(\mu) d\mu = \underbrace{\frac{1}{\sqrt{2\pi} \sqrt{\sigma^2 + \tau^2}} e^{-\frac{(x-\beta)^2}{2(\sigma^2 + \tau^2)}}}_{N(\beta, \sigma^2 + \tau^2)}$$

# Mixture of Discrete & Continuous

Example:

$$\mu = \begin{cases} \mu_0 & p \\ \text{Unif}(-\frac{L}{2}, \frac{L}{2}) & 1-p \end{cases}$$

$$\bar{x} | \mu \sim N(\mu, \frac{\sigma^2}{n})$$

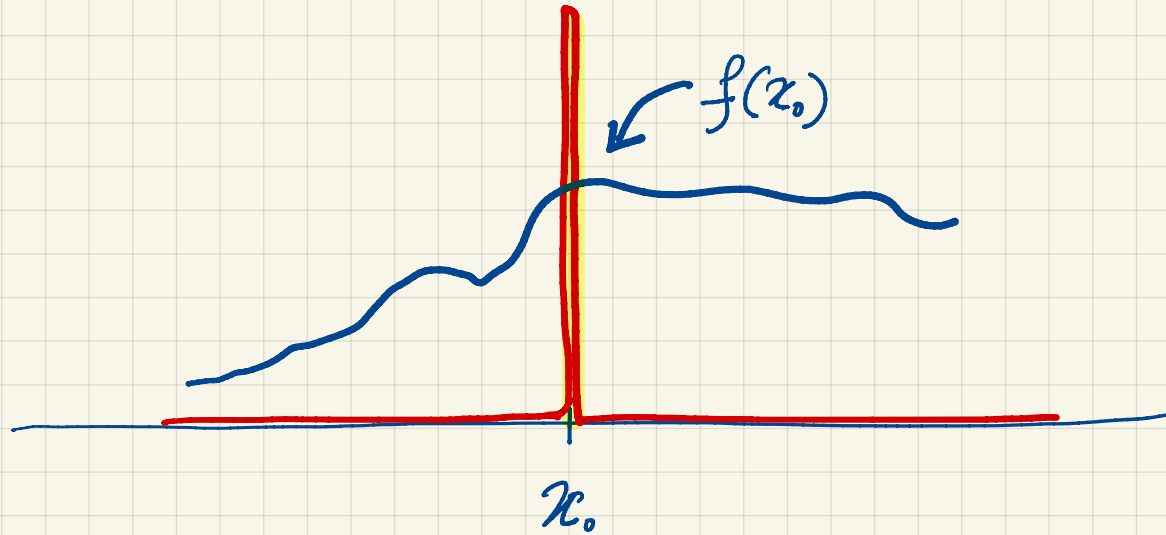


$$g(\mu) = \delta(\mu - \mu_0) \text{ where } \delta(x) = \begin{cases} \infty & x=0 \\ 0 & x \neq 0 \end{cases}$$

$$\text{Handle discrete with density: } \int_{-\infty}^{+\infty} \delta(x - x_0) dx = 1.$$

$$\text{More generally: } \int_{\mathbb{R}} \delta(x - x_0) f(x) dx = \begin{cases} f(x_0) & x_0 \in \mathbb{R} \\ 0 & \text{otherwise} \end{cases}$$

Replace  $x$  by  $x_0$  in  $f(x)$



$$\text{Area under } \delta(x - x_0) f(x) = 1 \cdot f(x_0) = f(x_0)$$

Is it consistent?

$$f(\mu) = p\delta(\mu - \mu_0) + (1-p)\frac{1}{L} \quad -\frac{L}{2} \leq \mu \leq \frac{L}{2}$$

$$P(\mu = \mu_0) = \int_{\mu_0}^{\mu_0} f(\mu) d\mu = \int_{\mu_0}^{\mu_0} p\delta(\mu - \mu_0) d\mu + \int_{\mu_0}^{\mu_0} p\frac{1}{L} d\mu$$

$$= p + 0 = p \quad \checkmark$$

Example: Coin with  $p(H) = p$

$$f(x) = p\delta(x-1) + (1-p)\delta(x)$$

$$P(x=0) = \int_0^0 f(x) dx = \int_0^0 p\delta(x-1) dx + \int_0^0 (1-p)\delta(x) dx = 0 + 1-p$$

$$P(x=1) = \int_1^1 f(x) dx = p$$

$$P(x=a) = \int_a^a f(x) dx = \int_a^a [p\delta(x-1) + (1-p)\delta(x)] dx = 0 \text{ if } 1 \notin [a, a] \text{ and } 0 \notin [a, a]$$



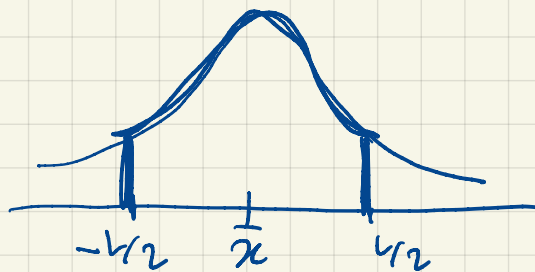
What is posterior due to  $g(\mu)$ ? Same! [Nothing can change this prior]

$$g(\mu|\bar{x}) = \frac{f(\bar{x}|\mu)g(\mu)}{\int_{-\infty}^{+\infty} f(\bar{x}|\mu)g(\mu)d\mu} = \frac{f(\bar{x}|\mu)\delta(\mu-\mu_0)}{\int_{-\infty}^{+\infty} f(\bar{x}|\mu)\delta(\mu-\mu_0)d\mu}$$
$$= \frac{f(\bar{x}|\mu)\delta(\mu-\mu_0)}{f(\bar{x}|\mu_0)} = \begin{cases} 1 \cdot \delta(0) & \mu = \mu_0 \\ 0 & \text{otherwise} \end{cases} = \delta(\mu-\mu_0)$$

What is posterior due to  $h(\mu)$ ?

$$h(\mu|\bar{x}) = \frac{f(\bar{x}|\mu) \frac{1}{L}}{\int_{-\frac{L}{2}}^{\frac{L}{2}} f(\bar{x}|\mu) \frac{1}{L} d\mu} \quad (\text{scaled Normal in } [-\frac{L}{2}, \frac{L}{2}])$$

Denominator is:  $\Phi\left(\frac{\frac{L}{2} - \bar{x}}{\sigma}\right) - \Phi\left(\frac{-\frac{L}{2} - \bar{x}}{\sigma}\right)$



Lindley's Paradox: (say that  $\mu_0$  is most likely no matter what!)

$$\frac{\alpha(\bar{x})}{\beta(\bar{x})} = \frac{\alpha(\bar{x})}{1 - \alpha(\bar{x})} = \frac{p \int_{-\infty}^{+\infty} f(\bar{x}|\mu) \delta(\mu - \mu_0) d\mu}{(1-p) \int_{-\frac{L}{2}}^{\frac{L}{2}} f(\bar{x}|\mu) \frac{1}{L} d\mu}$$

$$\geq \frac{p \int_{-\infty}^{+\infty} f(\bar{x}|\mu) \delta(\mu - \mu_0) d\mu}{(1-p) \int_{-\infty}^{+\infty} f(\bar{x}|\mu) \frac{1}{L} d\mu} = \frac{Lp}{1-p} f(\bar{x}|\mu_0)$$

Let  $\bar{x} = \mu_0 + c\sigma/\sqrt{n}$ . Then  $f(\bar{x}|\mu_0) = \frac{\sqrt{n}}{\sqrt{2\pi}\sigma} e^{-c^2/2}$

$$\frac{\alpha(\bar{x})}{1 - \alpha(\bar{x})} = \frac{Lp}{1-p} \frac{\sqrt{n}}{\sqrt{2\pi}\sigma} e^{-c^2/2} \rightarrow \infty \text{ when } n \rightarrow \infty$$

$\lim_{n \rightarrow \infty} \alpha(\mu_0 + c\sigma/\sqrt{n}) = 1$ , so posterior  $\approx \delta(\mu - \mu_0)$

Paradox: We observe  $\bar{x}$   $c$  s.d away from its mean  $\mu_0$ , but  $P(\mu = \mu_0 | \bar{x}) = 1$

Another approach: handle prob. & density in Bayes.

$$P(\mu = \mu_0 | \bar{x}) = \frac{f(\bar{x} | \mu = \mu_0) P(\mu = \mu_0)}{f(\bar{x} | \mu = \mu_0) P(\mu = \mu_0) + f(\bar{x} | \mu \neq \mu_0) P(\mu \neq \mu_0)}$$

$$\text{Now: } f(\bar{x} | \mu \neq \mu_0) = \int_{-\infty}^{\infty} f(\bar{x} | \mu) f(\mu) d\mu$$

$$P(\mu = \mu_0 | \bar{x}) = \frac{p f(\bar{x} | \mu = \mu_0)}{p f(\bar{x} | \mu = \mu_0) + (1-p) \int_{-\infty}^{\infty} f(\bar{x} | \mu) \frac{1}{L} d\mu}$$
$$\geq \frac{p f(\bar{x} | \mu = \mu_0)}{p f(\bar{x} | \mu = \mu_0) + \frac{1-p}{L}}$$

$$\bar{x} = \mu_0 + c\sigma/\sqrt{n} \Rightarrow P(\mu = \mu_0 | \bar{x}) \geq \frac{p \frac{\sqrt{n}}{\sqrt{2\pi}\sigma} e^{-c^2/2}}{p \frac{\sqrt{n}}{\sqrt{2\pi}\sigma} e^{-c^2/2} + \frac{1-p}{L}} \rightarrow 1$$

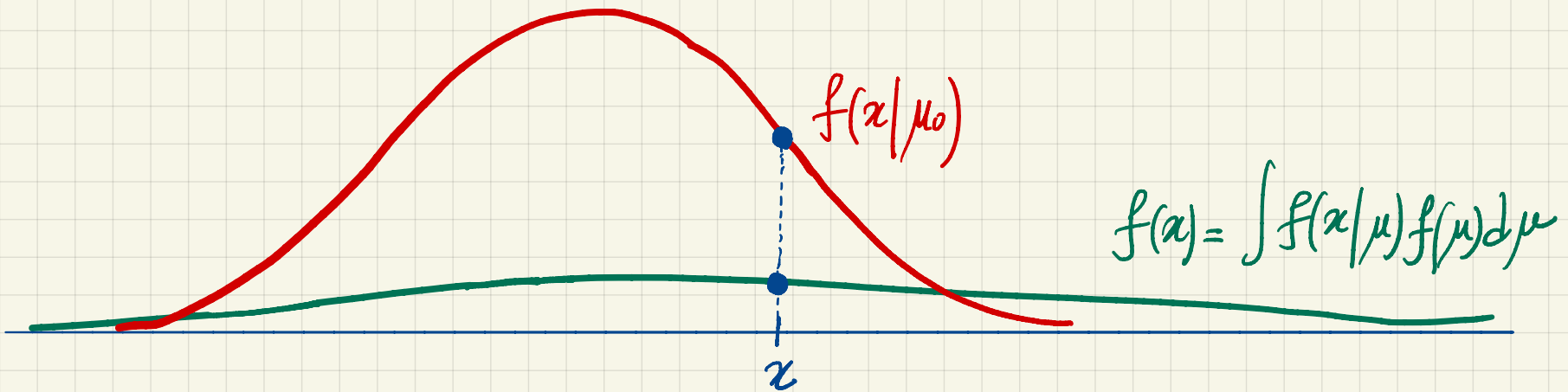
# Interpretation of Paradox

$x$  is observed with variance  $\sigma^2$

$\mu$  has distribution with variance  $\tau^2$

$$\sigma^2 \ll \tau^2$$

(in our example  $\frac{\sigma^2}{n} \rightarrow 0$ )



Ratio:  $\frac{f(x|\mu_0)}{f(x) = \int f(x|\mu) f(\mu) d\mu}$  becomes large.