

Chapter 7. Bayesian model:

① Review of conditional distribution:

If (x, y) have joint distribution $f(x, y)$, and x has marginal distribution $f_x(x)$. Then the conditional distribution of y given $X=x$ is

$$f(y|x) = \frac{f(x, y)}{f_x(x)}.$$

Eg1. If the joint pdf of (X, Y) is

$$f(x, y) = \begin{cases} 4xy & 0 \leq x, y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Find $f(y|x)$ for any $0 \leq x \leq 1$.

Sol: First find the marginal pdf of X . For $0 \leq x \leq 1$.

$$f_x(x) = \int_0^1 f(x, y) dy = \int_0^1 4xy dy = 2x.$$

$$\text{so. } f(y|x) = \frac{f(x, y)}{f_x(x)} = \begin{cases} 2y & 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Ex 2. If (x, y) have joint pdf

$$f(x, y) = \begin{cases} 8xy & \text{if } 0 \leq y \leq x \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

Find $f(y|x)$ for any $0 \leq x \leq 1$.

$$\text{Sol: } f_x(x) = \begin{cases} 4x^3 & \text{if } 0 \leq x \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

$$\text{so. } f(y|x) = \begin{cases} 2x^{-2}y & \text{if } 0 \leq y \leq x, \\ 0 & \text{otherwise.} \end{cases}$$

Ex 1. If (x, y) have joint pdf

$$f(x, y) = \begin{cases} x+y & \text{if } 0 \leq x, y \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

Find $f(y|x)$ for any $0 \leq x \leq 1$.

$$\text{Sol: } f_x(x) = \begin{cases} \frac{1}{2}(1+x) & \text{if } 0 \leq x \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

$$\text{so. } f(y|x) = \begin{cases} \frac{2(x+y)}{1+x} & \text{if } 0 \leq y \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

On the other hand, if we know the conditional distribution of y given $x=x$, i.e., $f(y|x)$. and also the marginal distribution of $f_x(x)$. Then the joint distribution of (x, y) is

$$f(x, y) = f(y|x) f_x(x).$$

Ex 3. If $f(y|x)$ is given below.

$$f(y|x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2}}. \quad \text{and} \quad f_x(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}.$$

then the joint pdf of (x, y) is

$$\begin{aligned} f(x, y) &= f(y|x) f_x(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \\ &= \frac{1}{2\pi} e^{-\frac{x^2+(y-x)^2}{2}}. \end{aligned}$$

Ex 2. If $f(y|x)$ is given by

$$f(y|x) = \begin{cases} \alpha e^{-(y-x)\alpha}, & y \geq x \\ 0. & \text{else.} \end{cases}$$

$$f_x(x) = \begin{cases} \alpha e^{-\alpha x}, & x \geq 0. \\ 0. & x < 0. \end{cases} \quad \text{find } f(x, y).$$

Sol:

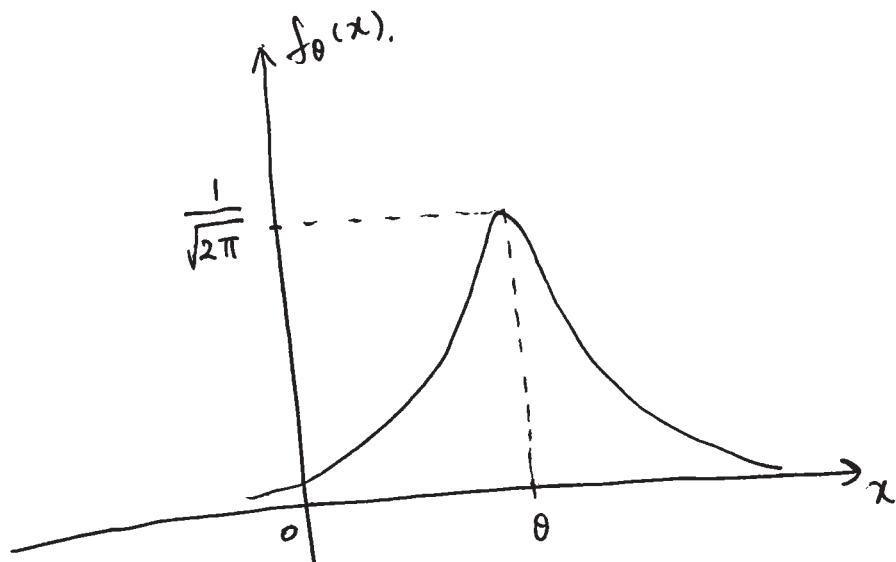
$$f(x, y) = \begin{cases} \alpha^2 e^{-\alpha y}, & \text{if } y \geq x \geq 0 \\ 0, & \text{else.} \end{cases}$$

② Bayesian model: Eg 5. Suppose $X \sim N(\theta, 1)$. That is, the pdf of

X is

$$f_\theta(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\theta)^2}{2}}, \quad x \in \mathbb{R}.$$

The plot of $f_\theta(x)$ is as below:



We don't know the value of θ .

In previous sections and chapters, we assume θ is deterministic but unknown, we proposed methods to estimate θ . e.g. use $\hat{\theta}_{ML}$, the MLE of θ . We all agree with this method.

The Bayesian statisticians don't agree with this. They believe that, since θ is unknown, θ can be anywhere on the real line so it should be treated as a random variable. People should even place a distribution on θ , i.e. the so-called prior distribution.

In this chapter, we introduce this somewhat new perspective.

Let us go back to our example. Bayesians believe we should ~~not~~ treat the location θ as a random variable, and they assume θ follows the following pdf $\theta \sim N(0, 1)$. i.e. the pdf of θ is

$$f(\theta) = \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}}.$$

Using this idea, the ~~junk~~ $f_\theta(x)$ should be considered as the conditional distribution of X given θ , i.e.

$$f(x|\theta) = f_\theta(x).$$

The joint pdf of (X, θ) becomes

$$\begin{aligned} f(x, \theta) &= f(x|\theta) f(\theta) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\theta)^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} \\ &= \frac{1}{2\pi} e^{-\frac{\theta^2+(x-\theta)^2}{2}} \cdot \cancel{e^{\frac{1}{2\pi}}}. \end{aligned}$$

Now we are able to find the conditional distribution of θ given $X=x$.

The marginal pdf of X is

$$f_X(x) = \int_{-\infty}^{\infty} f(x, \theta) d\theta = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{\theta^2 + (x-\theta)^2}{2}} d\theta$$

$$= \frac{1}{2\sqrt{\pi}} e^{-\frac{x^2}{4}}.$$

$$\text{so. } f(\theta|x) = \frac{f(x, \theta)}{f_X(x)} = \frac{\frac{1}{2\pi} e^{-\frac{\theta^2 + (x-\theta)^2}{2}}}{\frac{1}{2\sqrt{\pi}} e^{-\frac{x^2}{4}}} = \frac{1}{\sqrt{\pi}} e^{-(\theta - \frac{1}{2}x)^2}.$$

~~if $\theta \sim N(\mu, \sigma^2)$~~

$$\text{so } \theta|x=x \sim N\left(\frac{1}{2}x, \frac{1}{2}\right).$$

Change to the following motivating example:

Eg 4. If ~~x~~ $x \sim \text{Exponential}(\theta)$, that is, $f_\theta(x)$ is

$$f_\theta(x) = \begin{cases} \theta e^{-x\theta}, & x \geq 0. \\ 0, & \text{o.w.} \end{cases}$$

Assume θ follows $\theta \sim \text{Exponential}(1)$.

$$f(\theta) = \begin{cases} e^{-\theta}, & \theta \geq 0 \\ 0, & \text{o.w.} \end{cases}$$

then the joint pdf of (X, θ) is

$$f(x, \theta) = f(x|\theta) f(\theta)$$
$$= \begin{cases} \theta e^{-(x+1)\theta}, & \theta, x \geq 0. \\ 0, & \text{o.w.} \end{cases}$$

so the marginal pdf of X is

$$f_X(x) = \int_0^\infty f(x, \theta) d\theta = \int_0^\infty \theta e^{-(x+1)\theta} d\theta$$
$$= -\frac{1}{x+1} e^{-(x+1)\theta} \Big|_0^\infty + \int_0^\infty \frac{1}{x+1} e^{-(x+1)\theta} d\theta$$
$$= -\frac{1}{(x+1)^2} e^{-(x+1)\theta} \Big|_0^\infty = \frac{1}{(x+1)^2}.$$

so. $f(\theta|x) = \begin{cases} (x+1)^2 \theta e^{-(x+1)\theta}, & \theta \geq 0. \\ 0, & \text{o.w.} \end{cases}$

In summary. Suppose we know the distribution of X , i.e., $f_\theta(x)$.

Step 1. Treat $f_\theta(x)$ as the conditional distribution of X given θ , i.e.

$$f(x|\theta) = f_\theta(x).$$

Step 2. place a prior distribution $f(\theta)$ on θ .

Step 3. find joint distribution of (X, θ) by

$$f(x, \theta) = f(x|\theta) f(\theta) = f_\theta(x) f(\theta).$$

Step 4. Find the ^{posterior} ~~conditional~~ distribution ~~of~~ of θ given x ~~by~~.

$$f(\theta|x) = \frac{f(x, \theta)}{f_x(x)} = \frac{f(x, \theta)}{\int f(x, \theta) d\theta}, \text{ or simply, by } \\ f(\theta|x) \propto f(x, \theta) \\ = f_\theta(x) f(\theta).$$

also we can use the following example.

Eg 6. Suppose $X | \theta \sim \text{unif}(0, \theta)$. ~~$\theta \sim \text{unif}(0, 1)$~~ . θ has ^{prior} ~~prior~~

$$f(\theta) = \begin{cases} 2\theta, & 0 \leq \theta \leq 1, \\ 0, & \text{o.w.} \end{cases}$$

Find the posterior of θ given x .

Then $f(x|\theta) = \begin{cases} \frac{1}{\theta}, & 0 \leq x \leq \theta, \\ 0, & \text{o.w.} \end{cases}$ so $f(x|\theta) = \frac{1}{\theta} \mathbb{1}_{0 \leq x \leq \theta}$.

$$\text{so } f(x, \theta) = f(x|\theta) f(\theta) = \begin{cases} 2 \mathbb{1}_{0 \leq x \leq \theta}, & 0 \leq x \leq \theta \leq 1, \\ 0, & \text{o.w.} \end{cases}$$

$$\text{So } f_x(x) = \int_x^1 f(x, \theta) d\theta = \int_x^1 2 d\theta = 2(1-x).$$

$$\text{So } f(\theta|x) = \begin{cases} 2(1-x), & x \leq \theta \leq 1, \\ 0, & \text{otherwise.} \end{cases}$$