

## STAT 305 2024S2

### Lecture 6.2

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### Learning Goals

G1. Bayesian prior to posterior update

### General formula for Bayesian prior to posterior update

**Theorem 6.2 of course notes:**

Let observable  $Y$  be a RV with pmf/pdf  $f_{Y|\Theta}(y|\theta)$ , and  $\Theta$  a RV representing our uncertainty about the model parameter value  $\theta$ . Given prior distribution  $p_{\Theta}(\theta)$  for  $\Theta$ , the conditional/posterior distribution of  $\theta$  given  $Y$  takes value  $y$  is

$$p_{\Theta|Y}(\theta|y) = \frac{f_{Y|\Theta}(y|\theta)p_{\Theta}(\theta)}{f_Y(y)}$$

where

$$f_Y(y) = \begin{cases} \sum_{\theta} f_{Y|\Theta}(y|\theta)p_{\Theta}(\theta) & (\text{if } \Theta \text{ is discrete}) \\ \int f_{Y|\Theta}(y|\theta)p_{\Theta}(\theta)d\theta & (\text{if } \Theta \text{ is continuous}), \end{cases}$$

and the summation or integration is over all values  $\theta$  with  $p_{\Theta}(\theta) > 0$ .

Let's examine a couple of things:

- $p_{\Theta}(\theta)$  is a
- $f_{Y|\Theta}(y|\theta)$  is a
- $f_Y(y)$  is
- Posterior distribution of  $\Theta$  given  $Y = y$  is

What is the form for the posterior distribution for observing  $Y_1 = y_1, \dots, Y_n = y_n$  if the joint likelihood depends on a parameter  $\theta$ ?

### Example: small sample Bayesian inference

Let  $X_i$  represent the lifespan (measured in years) of the  $i$ -th sampled specimen of a bird of the species *Turdus philomelos* (Song Thrush). Goal is to infer average lifespan.

- We assume  $X_i, i = 1, \dots, 3$  to be i.i.d. samples from  $\mathcal{E}(\lambda)$  distribution. Let  $\theta = \frac{1}{\lambda}$ . We will use the likelihood function written in terms of  $\theta$  parameterization.
- We model uncertainty with regards to  $\theta$  via RV  $\Theta$  with prior distribution

$$p_{\Theta}(\theta) = \frac{\exp\left(-\frac{1}{2(1.5)^2}(\theta - 3)^2\right) \mathbb{1}_{(0,\infty)}(\theta)}{\int_0^{\infty} \exp\left(-\frac{1}{2(1.5)^2}(\theta - 3)^2\right) d\theta},$$

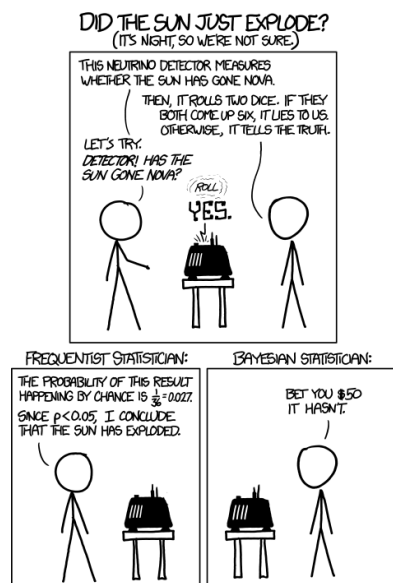
noting

$$\mathbb{1}_{(0,\infty)}(\theta) = \begin{cases} 1 & \text{if } \theta \in (0, \infty) \\ 0 & \text{otherwise.} \end{cases}$$

- Visualize the prior  $p_{\Theta}(\theta)$
- Write down the posterior distribution of  $\Theta$  given  $X_1 = 3, X_2 = 7, X_3 = 3.2$ , then visualize this posterior distribution.

## Lecture 6.2 Student Activities

1. Explain the following XKCD comic strip in terms of Bayesian prior to posterior update.  
(Source: <https://xkcd.com/1132/>)



2. A typical Bayesian application question. We will use Bayesian inference/learning to update our view about the proportion of UBC students who use their u-pass more than 3 days a week during the Winter terms in light of survey data. The survey consists of  $n$  randomly sampled UBC students, and let

$$X_i = \begin{cases} 1 & \text{if student uses uPass on more than 3 days a week on average} \\ 0 & \text{otherwise,} \end{cases}$$

and are assumed to be i.i.d. for all  $i = 1, \dots, n$ .

- (a) State your data model and reasoning.
- (b) What are you trying to learn in terms of the data-model parameter?
- (c) Write down the (joint)-likelihood for the data given parameter.
- (d) Write down your prior distribution for the model parameter and justify your choice, including hyper-parameters. You should choose one of the distributions in Table 1.4.
- (e) Write down the posterior distribution of the model parameter given data? (Simplify it as much as possible... but we will learn more about techniques in Lecture 6.3.)