## PSTAT 115 - Section Three

Winter 2023

## 1 Summary of Model

Review: What are the 4 ingredients common to each and every Bayesian model?

- i) *Prior Model:* We have a variable of interest  $\pi$ . The prior model specifies two pieces of info, 1. The values that  $\pi$  takes on and 2. the plant birty of each value.
- ii) Data Dependence: to learn about  $\pi$ , we take collect data Y (another random variable). Summarize the depondence of the data on  $\pi$  with **conditional pmf**.  $89\pi$
- iii) Likelihood function: after observing the data (i.e.  $\underline{Y=g}$ ), we have  $\underline{\chi}(y|\pi)$ . This represents the likelihood of seeing the dataset y given  $\underline{\lambda}(y|\pi)$  values (of  $\pi$ ).

**Question 1.** Let  $K = \int L(\lambda; y) p(\lambda) d\lambda$  be the integral of the proportional posterior. Then the proper posterior density,

iv) Posterior model: aims to balance the prior & **likelihood**. We get,

$$\boxed{f(\pi|y)} = \frac{f(\pi)L(\pi|y)}{f(y)} \propto f(\pi)L(\pi|y) \Rightarrow \text{Posterior} = \text{Prior. likelihood} \ .$$

## 2 Gamma Integral

i.e. a true density integrates to 1, can be expressed as  $p(\lambda|y) = \frac{L(\lambda;y)p(\lambda)}{K}$ . Compute this posterior density & express the density as a mixture of two gamma distributions. Notes

**Question 1.** Let  $K = \int L(\lambda;y)p(\lambda)d\lambda$  be the integral of the proportional posterior. Then the proper posterior density, i.e. a true density integrates to 1, can be expressed as  $p(\lambda|y) = \frac{L(\lambda;y)p(\lambda)}{K}$ . Compute this posterior density & express the density as a mixture of two gamma distributions.

Note: Equation above, 
$$p(x|y) = \frac{L(x|y)p(x)}{g(y)}$$

= 
$$2\Gamma(x;3)b(x)gy$$
 =  $\Gamma$ 

Step 1) Find k.

- Equation. 
$$\Gamma(z) = \int_{0}^{\infty} \lambda^{z-1} e^{-\lambda} d\lambda$$
.

$$K = \int e^{-1767^{2}} \lambda^{8} (\frac{2000^{3}}{\Gamma(3)} \lambda^{2} e^{-3000\lambda} + \frac{1000^{7}}{\Gamma(7)} \lambda^{6} e^{-1000\lambda}) d\lambda$$

$$= S\left(e^{-1767\lambda} \cdot 2000^{3} \cdot \lambda^{10} e^{-2000\lambda} + 1000^{7} e^{-1767\lambda} \frac{\lambda^{14} e^{-1000\lambda}}{\Gamma(7)}\right) d\lambda$$

= 
$$\frac{2000^3}{\Gamma(3)}$$
 Se-17672.  $\lambda^{10}$ .  $e^{-2000\lambda} + \frac{1000^7}{\Gamma(7)}$  S $\lambda^{14}$ e-17672  $e^{-1000\lambda}d\lambda$ 

$$= \frac{2000^{3}}{\Gamma(3)} \frac{5^{10}e^{-3767^{3}}d\lambda + 1000^{7}}{\Gamma(7)} \frac{5^{11}e^{-2767^{3}}d\lambda}{1000^{7}}$$

T1: 5 210e-37672 d2

$$= \frac{1}{3767} \int e^{-u} \cdot \left(\frac{1}{3767}\right)^{10} 0^{10} d0$$

$$=\frac{\Gamma(15)}{(2767)}$$
15.

Combining T1, T2.

$$K = \frac{2000^3}{\Gamma(3)} \cdot \frac{\Gamma(11)}{(3767)''} + \frac{1000^7}{\Gamma(7)} \cdot \frac{\Gamma(15)}{(2767)^{15}}$$

· Step 2) compute p(xly)

where K=SL(x; x)p(x)dx.

Answer: 
$$\frac{2000^3}{\Gamma(3)} \lambda^{10} e^{-3767\lambda} + \frac{1000^7}{\Gamma(7)} \lambda^{14} e^{-2767\lambda}$$

Simplifies to,

$$\frac{2000^{3}}{\Gamma(3)} \cdot \frac{\Gamma(11)}{3767^{11}} = \frac{2767^{11}}{\Gamma(11)} \times \frac{2000^{3}}{\Gamma(11)} \times \frac{\Gamma(15)}{2767^{15}} \times \frac{2767^{11}}{\Gamma(3)} \times \frac{2767^{11}}{3767^{15}} \times \frac{2767^{15}}{3767^{15}} \times \frac{2767^{15}}{3767^$$

$$= kp_{\omega}(x) + (1-\omega)p_{\omega}(x)$$

i.e. posterior density mixture & a gamma distributions