

**HW 2 – 4803, Fall 2024**  
**Each problem is worth 10 points**

## 1 Part I – theoretical problems

1. Suppose we have a dataset that consists of  $n$  observations of the outcomes  $y_i$  of a Poisson process. For such a process,  $y_i$  must be a non-negative integer, and the probability distribution for any given  $y$  is the Poisson distribution

$$P(y|\lambda) = \frac{\lambda^y e^{-\lambda}}{y!} ,$$

where  $\lambda > 0$ .

Now suppose we want to estimate what the parameter  $\lambda$  from the above equation is for the Poisson process that generated our data.

- (a) Show that the conjugate prior to the Poisson distribution above is the Gamma distribution with hyperparameters  $\alpha$  and  $\beta$ :

$$P(\lambda) = \frac{\lambda^{\alpha-1} e^{-\beta\lambda} \beta^\alpha}{\Gamma(\alpha)} ,$$

where  $\Gamma(x)$  is the gamma function (but this is not so important here). That is, if we choose our prior for  $\lambda$  to be the Gamma distribution, then the posterior for  $\lambda$  will also be a Gamma distribution with some new hyperparameters  $\alpha'$  and  $\beta'$ . Find formulas for  $\alpha'$  and  $\beta'$ .

- (b) Suppose after forming our posterior  $P(\lambda|y)$  we are then asked to predict the outcome  $z$  of another observation from this same Poisson process. Explain why a logical way to do so is via the *posterior predictive* distribution:

$$P(z|y) = \int_0^\infty P(z|\lambda) P(\lambda|y) d\lambda .$$

Compute the posterior predictive distribution in this case (hint: it is a negative binomial distribution).

- 2-5. From the Book “An Introduction to Statistical Learning” – 4.8 Exercises 6,7,8,12
- 6-8. From the Book “The Elements of Statistical Learning” – Chapter 3 Exercises 3.12, 3.28, and 3.29. These problems might be challenging!

## **2 Part II – programming**

**Programming Problems 1-2** From the Book “An Introduction to Statistical Learning” – 4.8 Exercises 14 and 16.