

# Homework 2

## Bayesian Data Analysis

### Problem 1

Suppose you are interested in estimating the average total snowfall per year  $\mu$  (in inches) for a large city on the East Coast of the United States.

Before you collect the data, you are 95% confident that the average snowfall could be between 38 and 42 inches, centered around 40 inches per year.

Assume individual yearly snow totals  $y_1, \dots, y_n$  are collected from a population that is assumed to be normally distributed with mean  $\mu$  and known standard deviation  $\sigma = 10$  inches.

Suppose you observe the yearly snowfall totals 38.6, 42.4, 57.5, 40.5, 51.7, 67.1, 33.4, 60.9, 64.1, 40.1, 40.7, and 6.4 inches.

- (1) Propose a conjugate Normal prior  $\mu \sim N(\mu_0, \tau_0)$  for this problem.
- (2) What would be the posterior distribution in such case? What can you comment about the posterior mean, and the relative contribution of prior and data to the posterior expectation?
- (3) Plot the prior and the posterior distribution in a single figure. What can you comment about the posterior distribution of  $\mu$ ?
- (4) Compute the 10% and 90% quantiles of the posterior distribution (i.e. those values  $a$  and  $b$  such that  $P(\mu < a | \text{Data}) = 0.1$  and  $P(\mu < b | \text{Data}) = 0.9$ , respectively).
- (5) For reporting needs, instead of computing the posterior mean of  $\mu$  you are asked to compute the posterior mean of  $\log(\mu)$ . How would you compute it?

### Problem 2

The Mayo Clinic conducted a study of  $n = 50$  patients followed for one year on a new medication and found that 30 patients experienced no adverse side effects (ASE), 12 experienced one ASE, 6 experienced two ASEs and 2 experienced ten ASEs.

- (1) Derive the posterior of  $\lambda$  for the Poisson/gamma model  $Y_1, \dots, Y_n | \lambda \stackrel{\text{id}}{\sim} \text{Poisson}(\lambda)$  and  $\lambda \sim \text{Gamma}(a, b)$ .
- (2) Use the Poisson/gamma model with  $a = b = 0.01$  to study the rate of adverse events. Plot the posterior and give the posterior mean and 95% credible interval.

- (3) In Bayesian Analysis, it is important to assess the *sensitivity* of the posterior inference to prior specifications. Repeat the analysis with Gamma(0.1,0.1) and Gamma(1,1) priors and discuss the sensitivity of the results to the prior.
- (4) Plot the data versus the Poisson(  $\hat{\lambda}$  ) PMF, where  $\hat{\lambda}$  is the posterior mean of  $\lambda$  from part (2). Does the Poisson likelihood fit the data well?
- (5) The current medication is thought to have around one adverse side effect per year. What is the posterior probability that this new medication has a higher side effect rate than the previous medication? Are the results sensitive to the prior?

### Problem 3

Over the past 50 years California has experienced an average of  $\lambda_0 = 75$  large wildfires per year. For the next 10 years you will record the number of large fires in California and then fit a Poisson/gamma model to these data. Let the rate of large fires in this future period,  $\lambda$ , have prior  $\lambda \sim \text{Gamma}(a, b)$ . Select  $a$  and  $b$  so that the prior is somehow uninformative, i.e. has prior variance around 100 and gives prior probability approximately  $\text{Prob}(\lambda > \lambda_0) = 0.5$ . This prior assessment corresponds to an equal probability a priori assigned to the hypothesis that the rate of fires may increase in the next 10 years *vs* that it may decrease.