

# Bayesian Modeling Problems

ELG 5218 – Uncertainty Evaluation in Engineering Measurements and Machine Learning

Instructor: Miodrag Bolić, University of Ottawa

Date: January 28, 2026

## PART A: CONCEPTUAL QUESTIONS

A1. What is the difference between a 95% frequentist *confidence interval* and a 95% Bayesian *credible interval* for a parameter?

A2. Name and briefly describe the two types of uncertainty often characterized in Bayesian modeling.

A3. What is a *conjugate prior*? Give an example of a conjugate prior-likelihood pair.

A4. Explain the distinction between Maximum Likelihood Estimation (MLE) and Maximum A Posteriori (MAP) estimation for a model parameter.

## PART B: MATHEMATICAL DERIVATIONS

B1. Consider a Binomial likelihood with  $k$  successes out of  $N$  trials for an unknown probability  $\theta$ , and a  $\text{Beta}(\alpha, \beta)$  prior for  $\theta$ .

- Derive the posterior distribution for  $\theta$  (show that the Beta prior is conjugate to the Binomial likelihood).
- Find the MAP (Maximum A Posteriori) estimate of  $\theta$  from this posterior, and compare it to the MLE (Maximum Likelihood Estimate).

## PART C: PARAMETRIC ANALYSIS

C1. You consider two Bayesian models for a probability  $\theta$  (e.g. the fraction of defective items in manufacturing) with different prior strengths:

- Model A:  $\theta \sim \text{Beta}(1, 1)$  (a very weak, uniform prior).
- Model B:  $\theta \sim \text{Beta}(10, 10)$  (a stronger prior peaked around 0.5).

Both models are updated on the same dataset. Answer the following:

- Which model's posterior distribution for  $\theta$  will have the larger variance? Explain why.
- Which model is likely to yield more extreme posterior *predictive* probabilities for new observations (i.e. predictions closer to 0 or 1)? Why?
- In a scenario where  $\theta$  is actually very small (a rare event case), which prior (Model A or Model B) would be more appropriate to use, and why?

## PART D: IMPLEMENTATION AND PRACTICAL QUESTIONS

D1. In a Beta-Binomial model, you have updated the posterior for  $\theta$  given observed data. How can you compute the probability that the next trial will be a success? Provide:

- an analytic expression for this posterior predictive probability in terms of the posterior Beta parameters, and
- a brief description of how you could approximate this probability via Monte Carlo simulation.

PART E: OTHER PROBLEMS

E1. What is an *equal-tailed* 95% credible interval, and how does it differ from a 95% *highest posterior density* (HPD) credible interval? In what situations would these two types of intervals give notably different results?

E2. **Problem: Manufacturing Defect Detection.** A manufacturing line produces electronic components, and the historical defect rate is around 5%. You decide to use Bayesian modeling to monitor the defect rate in production. You assume a Beta prior for the defect probability  $\theta$ . In a random sample of  $N = 10$  components from a new batch, you observe  $X = 2$  defective units.

- (a) Why might a Bayesian approach be preferable to a frequentist approach (e.g. using just the sample proportion) for estimating  $\theta$  in this scenario?
- (b) Assume a prior  $\theta \sim \text{Beta}(2, 38)$  reflecting the historical belief (mean  $\approx 5\%$ ). Update this prior with the data. Give the posterior distribution for  $\theta$ , and calculate the posterior mean and a 95% credible interval for  $\theta$ .