

Worksheet 17

1. Consider a prior distribution $p \sim \text{Beta}(\alpha, \beta)$ for some fixed α and β for a likelihood given by $X|p \sim \text{Bin}(n, p)$. Derive the posterior distribution $p|X$.

2. For reasons that we will explore in more next time, the Bayesian point estimator (the best single-number estimator) is the expected value of the posterior distribution. Under the set up from the previous question, what the Bayesian point estimator \hat{p} in terms of X , α , and β ?

3. Consider observing $X \sim \text{Bin}(n, p)$. We know that the MLE estimator of p is given by $\hat{p}_{MLE} = X/n$. The Binomial comes from doing n Bernoulli trials and adding the number of 1s. Consider creating a new Y in which we artificially augment the data X by adding (in effect) an extra 0 and an extra 1. In other words, we create a $Y = X + 1$ with the assumption that $Y \sim \text{Bin}(n + 2, p)$. What is the MLE of p using the data from the augmented data Y ? Where have you seen this before?

4. Consider your solution to the previous set of questions. If α and β are non-negative integers, how could you describe the Bayesian estimator based on adding data to X ?

5. The standard uniform distribution is equivalent to $\text{Beta}(1, 1)$. In the notes I started by implicitly assuming that this was a fairly neutral starting assumption for indicating that we do not have any strong prior knowledge of p . Based on your results above, what would actually seem to be the best natural position if we do not want the prior to have a strong influence on the posterior mean?