Bayesian statistics

Lecture 3

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Find Jeffreys' prior for mean parameter in a Gaussian model with known variance σ^2 .

Find Jeffreys' prior for variance parameter in a Gaussian model with known mean μ .

Suppose we have samples $x_1, x_2, ..., x_5 = 0,0,1,1,0$ from a Bernoulli distribution with parameter $\theta \in (0,1)$.

a) Calculate the MLE.

Suppose we impose the restriction that $\theta \in \{0.2, 0.5, 0.7\}$.

b) Calculate the MLE.

Now we are going to be Bayesian. Suppose we have a prior $\pi(0.2) = 0.1, \pi(0.5 = 0.01), \pi(0.7) = 0.89$.

c) What is the MAP estimator?

Exercise 3 part 3

- d) Show that we can make the MAP whatever we like, by finding a prior over $\{0.2,0.5,0.7\}$ so that the MAP is 0.2, another so that it is 0.5, and another so that it is 0.7.
- e) Typically for the Bernoulli distribution, if we use MAP, we want to be able to get any value in [0,1], not just the ones in a finite set such as above. So we need a continuous prior distribution with range [0,1]. We assign the Bernoulli parameter a beta prior: $\theta \sim Beta(\alpha,\beta)$. Recall that the mode of a $Beta(\alpha,\beta)$ random variable is $(\alpha-1)/((\alpha-1)+(\beta-1))$. Suppose x^n are iid Bernoulli with unknown parameter. Show that the posterior has a $Beta(k+\alpha,n-k+\beta)$ distribution, and find the MAP estimator.
- f) If we use a Beta(1,1) a uniform distribution as prior, how would the MAP and MLE compare?

Exercise 3 part 3

- g) Since the posterior is also a Beta distribution, we call Beta the conjugate prior to the Bernoulli distribution. Interpret how the prior parameters α and β affect our MAP estimate.
- h) As the number of samples goes to infinity, what is the relationship between MLE and MAP? What does this say about our prior when n is small, or n is large?
- i) Would you prefer MLE or MAP? Why?

Credible intervals are not unique. Come up with a simple posterior distribution and show more than one valid 95% credible interval.