

Lab 7

Questions with a (*) next to them form part of homework 2. Due in 2/11/2015. [Q 4]

- Suppose $X_1, \dots, X_n \sim \text{Poisson}(\lambda)$ iidrv so $\mathbb{P}(X_i = k) = e^{-\lambda} \frac{\lambda^k}{k!}$.
 - What, roughly speaking, is the distribution of $\bar{X} = \sum_{i=1}^n X_i/n$? [i.e. use the CLT]
 - Given λ , find an interval such that \bar{X} belongs to that interval with probability 95%.
 - Re-arrange your answer to part (b) to produce an approximate 95% confidence interval for λ .
 - Let $\lambda = 2$ and $n = 10$. Using R, do the following a large number of times: Sample $X_1, \dots, X_n \sim \text{Poisson}(\lambda)$ and check if your confidence interval includes λ . What fraction of the time does your confidence interval include λ .
 - How about when $n = 100$?
 - Let $n = 1$ and assume λ is large. Argue that $\sqrt{\bar{X}} \approx N(\sqrt{\lambda}, 1)$. Use that approximation to form a confidence interval for $\sqrt{\lambda}$ and therefore for λ . [This is an example of a variance stabilising transform!]
- Suppose $X_1, \dots, X_n \sim N(\mu, \sigma^2)$ i.i.d.r.v. with unknown μ, σ^2 .
 - Produce 95% confidence intervals for μ and σ^2 in terms of \bar{X} and S^2 .
 - Use R to sample some data from the $N(1, 10^2)$ distribution and apply your answer from part (a).
- Let $X_1, \dots, X_n \sim \text{Uniform}(0, 1/\theta)$ where $\theta > 0$ is some unknown parameter; $X = (X_1, \dots, X_n)$. Taking the Bayesian route, take (θ, X) to have a joint distribution f . Take your prior distribution to be the distribution to be

$$f(\theta) = 1/\theta, \quad f(\theta) = \int f(\theta, x) dx.$$

- Is this a proper prior? i.e. is $\int f(\theta) d\theta = 1$? In words, what does our choice of prior say about our beliefs about θ ?
 - Use the formula
$$f(\theta | x) \propto f(\theta) \prod_{i=1}^n f(X_i | \theta)$$
to find the posterior distribution? Is it a proper distribution? i.e. is $\int f(\theta | x) d\theta = 1$?
 - What is the posterior mean? Does that seem reasonable?
- (*) Let $X_1, \dots, X_n \sim N(\theta, 1)$ and take your prior beliefs about θ be $N(a, b^2)$. You will observe X_1, \dots, X_n from the distribution $N(\theta, \sigma^2)$ where σ is known.
 - Find the posterior distribution. Try to express your answer as simply as possible.
 - Calculate the limit of the posterior distribution as $n \rightarrow \infty$ with a, b, σ fixed.
 - Let $a = 0, b = 10$. Setting $\theta = 100, \sigma = 1$ and $n = 8$, sample X_1, \dots, X_n . Calculate the posterior distribution. Find a 98% credible interval for θ by calculating the 1st and 99th quantiles of the posterior distribution using `qnorm`.
 - Repeat part (c) but $b = 1/10$. Comment on the difference.
 - An experiment will produce a result X with distribution $\text{Binomial}(100, p)$. A scientist tells you that their prior beliefs is that p is roughly between 0.7 and 0.8. Formulate a suitable prior distribution using the Beta family of distributions. You observe a sample $x = 65$.
 - Plot the prior distribution on the interval $[0, 1]$.
 - Calculate the posterior distribution and plot it. Compare with part (a).
 - Calculate a 95% posterior credible interval for p . [Hint: use `qbeta`] Does this seem to be compatible with your observation?
 - Repeat parts (b) and (c) but based on an observation of $x = 10$.

6. Let $X_1, \dots, X_n \sim \text{Poisson}(\lambda)$ where λ is unknown.

(a) Calculate the Jeffreys prior for the $\text{Poisson}(\lambda)$ distribution

$$\text{prior} \propto \sqrt{I(\theta)} \quad I(\theta) = \text{Var}_{X \sim \text{Poisson}(\theta)} \left(\frac{\partial}{\partial \theta} \log f(X | \theta) \right) \quad f(x | \theta) = e^{-\theta} \frac{\theta^x}{x!}$$

Calculate the posterior distribution with respect to the Jeffreys prior.

(b) After taking expert advice you decide instead that your prior beliefs about λ are given by a Gamma distribution with mean 100 and variance 200. Calculate the posterior distribution.