## APPM 4/5720: Computational Bayesian Statistics, Spring 2018

## Problem Set One (Due Friday, January 26th)

1. Let $X_{1}, X_{2}, \ldots, X_{n}$ be independent and identically distributed normal random variables with mean $\mu$ and variance $\sigma^{2}$. (We write $X_{1}, X_{2}, \ldots, X_{n} \stackrel{i i d}{\sim} N\left(\mu, \sigma^{2}\right)$.)
Assume that $\sigma^{2}$ is fixed and assume a $N\left(\mu_{0}, \sigma_{0}^{2}\right)$ prior distribution for the parameter $\mu$. (Assume that $\mu_{0}$ and $\sigma_{0}^{2}$ are known hyperparameters.)
Find the posterior distribution for $\mu$. (Name it.) Is the normal distribution a conjugate prior for $\mu$ for this model? Explain.
2. Suppose that $X$ has a gamma distribution with parameters $\alpha$ and $\beta$. We write $X \sim \Gamma(\alpha, \beta)$. In our class, this will always mean that $X$ has the pdf

$$
f(x)=\frac{1}{\Gamma(\alpha)} \beta^{\alpha} x^{\alpha-1} e^{-\beta x} I_{(0, \infty)}(x)
$$

(Whereas for some people it means that the pdf is $f(x)=\frac{1}{\Gamma(\alpha)} \frac{1}{\beta^{\alpha}} x^{\alpha-1} e^{-x / \beta} I_{(0, \infty)}(x)$.)
For our parameterization, $\alpha$ is known as a shape parameter and $\beta$ is known as an inverse scale parameter. (For the other parameterization, $\beta$ is known as the scale parameter.)
(a) Suppose that $X \sim \Gamma(\alpha, \beta)$. Find the probability density function for $Y:=1 / X$. Although this might not be a "recognizable" distribution for you, we call the distribution the "inverse gamma distribution" and will denote it as $Y \sim I G(\alpha, \beta)$.
(b) Suppose that $X_{1}, X_{2}, \ldots, X_{n} \stackrel{i i d}{\sim} N\left(\mu, \sigma^{2}\right)$ with $\mu$ fixed and known.

Assume an inverse gamma prior distribution for $\sigma^{2}$, with hyperparameters $\alpha$ and $\beta$. (i.e. that $\left.\sigma^{2} \sim I G(\alpha, \beta)\right)$
Find the posterior distribution for $\sigma^{2}$. Is the inverse gamma a conjugate prior for this normal model? Explain.
3. Let $X_{1}, X_{2}, \ldots, X_{n}$ be independent and identically distributed exponential random variables with rate $\lambda$. Find a conjugate prior for $\lambda$. (Prove that it is a conjugate prior.)
4. (Gelman et.al., Ch 1, Problem 3) Probability calculation for genetics (from Lindley, 1965): Suppose that in each individual of a large population there is a pair of genes, each of which can be either x or X , that controls eye color: those with xx have blue eyes, while heterozygotes (those with Xx or xX ) and those with XX have brown eyes. The proportion of blue-eyed individuals is $p^{2}$ and of heterozygotes is $2 p(1-p)$, where $0<p<1$. Each parent transmits one of its own genes to the child; if a parent is a heterozygote, the probability that it transmits the gene of type X is $\frac{1}{2}$. Assuming random mating, show that among brown-eyed children of brown-eyed parents, the expected proportion of heterozygotes is $2 p /(1+2 p)$. Suppose Judy, a brown-eyed child of brown-eyed parents, marries a heterozygote, and they have $n$ children, all brown-eyed. Find the posterior probability that Judy is a heterozygote and the probability that her first grandchild has blue eyes.
5. The random variables $X_{1}, X_{2}, \ldots, X_{n}$ are said to be exchangeable if the joint distribution of $\left(X_{1}, X_{2}, \ldots, X_{n}\right)$ is the same as the joint distribution of $\left(X_{\pi(1)}, X_{\pi(2)}, \ldots, X_{\pi(n)}\right)$ for any permutation $\pi$ of the indices $\{1,2, \ldots, n\}$.
(a) Show that if $X_{1}, X_{2}, \ldots, X_{n}$ are iid, then they are exchangeable.
(b) Give an example, ( $n=2$ is fine) to show that exchangeability does not necessarily imply "iid-ness".

