Name:

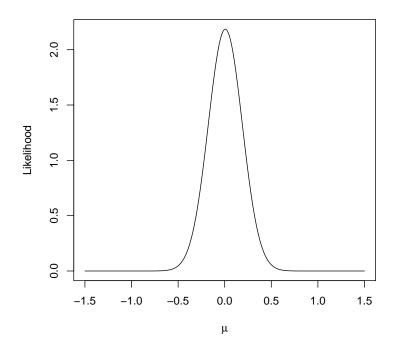
Midterm Examination October 9, 2013

- 1. Conjugate Priors.
 - (a) (3 pts) Define a conjugate prior.

(b) (4 pts) What is the conjugate prior for the Exponential(λ) sampling distribution? Use the sampling distribution $f(x \mid \lambda) = \lambda \exp(-\lambda x)$. Both state the conjugate prior and show that it is conjugate.

2. Jeffreys' Prior. (5 pts) Derive the Jeffreys' prior for the normal mean assuming that the variance is known.

3. **Priors**. The plot is the likelihood function for the mean of a Normal(μ , 1) distribution evaluated using 30 observations.



(a) (3 pts) Sketch (and label) a diffuse prior on the plot. Write a short definition of a diffuse prior.

(b) (3 pts) Sketch (and label) an *informative* prior on the plot. Write a short definition of an informative prior.

(c) **(4 pts)** Sketch (and label) the Jeffreys' prior on the plot (see Question 2). Define an *improper* prior. Is the Jeffreys' prior proper or improper?

4. Predictive Distribution.

(a) (3 pts) Let $X_1, \ldots, X_n \mid p \sim \text{Bernoulli}(p)$ and assume that we observe x successes and n-x failures. Let the prior distribution for p be $\text{Beta}(\alpha, \beta)$. What is the posterior distribution for p?

(b) (3 pts) Describe how to draw a sample from the predictive distribution of X_{n+1} assuming that you have a sample $p^{(i)}, i = 1, ..., m$ from the posterior distribution of p.

(c) (4 pts) Using the posterior distribution from (4a), derive the predictive distribution for X_{n+1} assuming that it is drawn from the same Bernoulli(p) distribution.

5. Sampling.

(a) **(5 pts)** Suppose that you have a posterior distribution $\pi(\theta \mid x_1, \ldots, x_n)$ with support on (0,1). Describe how you would use *brute force* sampling to draw a random sample from $\pi(\theta \mid x_1, \ldots, x_n)$.

(b) (3 pts) Suppose that we have a sample $\theta^{(i)}$, i = 1, ..., m from $\pi(\theta \mid x_1, ..., x_n)$. How could we use this sample to evaluate the posterior probability that $\theta > 0.8$?

6. Useful densities.

• Bernoulli distribution

$$f(x \mid p) = p^x (1-p)^{1-x}$$

• Beta distribution

$$f(x \mid \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$

• Binomial distribution

$$f(x \mid p) = \binom{n}{x} p^x (1-p)^{n-x}$$

• Exponential distribution

$$f(x \mid \lambda) = \lambda \exp(-\lambda x)$$

• Gamma distribution

$$f(x \mid \alpha, \beta) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} \exp(-\beta x) x^{\alpha - 1}$$

• Inverse gamma distribution

$$f(x \mid \alpha, \beta) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} \exp(-\beta/x) x^{-\alpha-1}$$

• Normal distribution

$$f(x \mid \mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp(-\frac{1}{2\sigma^2}(x - \mu)^2)$$

• Poisson distribution

$$f(x \mid \lambda) = \frac{\lambda^x \exp(-\lambda)}{x!}$$