Conceptual and Multiple Choice Exercises

Select the correct answer(s). There could be one answer or multiple correct answers for a problem.

- 1. What is the minimal condition that must be satisfied in order to use an improper prior in Bayesian inference?
 - (a) The posterior distribution must be continuous
 - (b) The prior distribution must be symmetric
 - (c) The prior distribution must have finite mean and variance
 - (d) The posterior distribution must be proper
- 2. Which of the following is not a meaningful prior elicitation approach?
 - (a) Expert opinion
 - (b) Prior centered weakly at the MLE
 - (c) Point mass at the MLE
 - (d) Flat uninformative prior
- 3. Which of the following is a model diagnostics tool?
 - (a) posterior predictive checks
 - (b) trace plots
 - (c) the ergodic theorem
 - (d) hypothesis testing
- 4. The effective sample size in an MCMC procedure can be interpreted as the number of dependent Monte Carlo samples necessary to give the same precision as the samples from MCMC.
 - (a) True
 - (b) False
- 5. If a conjugate prior is available for a sampling distribution (model), we should always adopt the conjugate prior for easy computation.
 - (a) True

- (b) False
- 6. Suppose θ^* is a new data point and θ^s is the current state of your Metropolis sampler. Which are symmetric proposal (jumping) distributions? Answer: B and C.
 - (a) $J(\theta^* \mid \theta^s) = Normal(0, \delta^2)$.
 - (b) $J(\theta^* \mid \theta^s) = Uniform(\theta^s \delta, \theta^s + \delta)$
 - (c) $J(\theta^* \mid \theta^s) = Normal(\theta^s, \delta^2)$.
 - (d) None of the above are symmetric.
- 7. The burn-in period in MCMC is theoretically justified.
 - (a) True
 - (b) False
- 8. We frequently report the posterior mean as our Bayes estimator because it
 - (a) minimizes the posterior expected linear loss
 - (b) minimizes the posterior expected quadratic loss
 - (c) maximizes the posterior expected linear loss
 - (d) maximizes the posterior expected quadratic loss
- 9. Let y_1, \ldots, y_n be a random iid sample from an exponential distribution, where

$$p(y_i \mid \theta) = \theta e^{-\theta y_i}.$$

Suppose the prior is

$$p(\theta) \propto \theta^{\alpha - 1} \exp{-\beta \theta}$$
.

Write the posterior mean as a weighted average of the prior mean and $\hat{\theta}$, which is an estimate of the parameter of θ that is only based upon the data. That is, write

$$E(\theta \mid y_1, \dots, y_n) = a \times \text{prior mean} + b \times \hat{\theta}.$$

What are a and b?

(a)
$$a = \frac{\beta}{\beta + \sum_{i} y_{i}}$$
 $b = \frac{\sum_{i} y_{i}}{\beta + \sum_{i} y_{i}}$

(b)
$$a = \frac{\alpha}{\alpha + \sum_{i} y_{i}}$$
 $b = \frac{\sum_{i} y_{i}}{\alpha + \sum_{i} y_{i}}$ (c) $a = \frac{\alpha}{\beta + \sum_{i} y_{i}}$ $b = \frac{\sum_{i} y_{i} + \alpha}{\beta + \sum_{i} y_{i}}$

(c)
$$a = \frac{\alpha}{\beta + \sum_{i} y_{i}}$$
 $b = \frac{\sum_{i} y_{i} + \alpha}{\beta + \sum_{i} y_{i}}$

(d)
$$a = \frac{\beta}{\beta + \alpha + \sum_{i} y_{i}}$$
 $b = \frac{\sum_{i} y_{i}}{\beta + \alpha + \sum_{i} y_{i}}$

- 10. Rejection sampling is fairly efficient as the dimension of the problem gets larger.
 - (a) True
 - (b) False