

Stat 352: Solutions to Assignment #1

1. Using a Riemann midpoint sum with four partitions of equal width gives

$$\int_0^1 x^2 dx \approx \frac{1}{4} \left[\left(\frac{1}{8}\right)^2 + \left(\frac{3}{8}\right)^2 + \left(\frac{5}{8}\right)^2 + \left(\frac{7}{8}\right)^2 \right] = \frac{21}{64} = 0.328125.$$

2. Note that a special case of Bayes' Theorem follows from the definition of conditional probability, namely that if $P(A) > 0$ and $P(B) > 0$, then $P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$ since both of these expressions equal $P(A \cap B)$. Hence,

$$\begin{aligned} & P(\text{Elvis was an identical twin} | \text{Elvis had a twin brother}) \\ &= \frac{P(\text{Elvis had a twin brother} | \text{Elvis was an identical twin}) \cdot P(\text{Elvis was an identical twin})}{P(\text{Elvis had a twin brother})}. \end{aligned}$$

We are told that

$$P(\text{Elvis was an identical twin}) = \frac{1}{300}.$$

Furthermore, it follows immediately that

$$P(\text{Elvis had a twin brother} | \text{Elvis was an identical twin}) = 1.$$

Therefore, we must calculate $P(\text{Elvis had a twin brother})$ using the law of total probability. Thus,

$$\begin{aligned} & P(\text{Elvis had a twin brother}) \\ &= P(\text{Elvis had a twin brother} | \text{Elvis was an identical twin}) \cdot P(\text{Elvis was an identical twin}) \\ &\quad + P(\text{Elvis had a twin brother} | \text{Elvis was a fraternal twin}) \cdot P(\text{Elvis was a fraternal twin}) \\ &\quad + P(\text{Elvis had a twin brother} | \text{Elvis was NOT a twin}) \cdot P(\text{Elvis was NOT a twin}) \\ &= 1 \cdot \frac{1}{300} + \frac{1}{2} \cdot \frac{1}{125} + 0 \end{aligned}$$

so that

$$P(\text{Elvis was an identical twin} | \text{Elvis had a twin brother}) = \frac{1 \cdot \frac{1}{300}}{1 \cdot \frac{1}{300} + \frac{1}{2} \cdot \frac{1}{125}} = \frac{5}{11}.$$

3. Suppose that $Y|\theta \sim \text{Bin}(n, \theta)$ with $\theta \sim \beta(a, b)$ so that

$$g(\theta) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \theta^{a-1} (1-\theta)^{b-1}, \quad 0 \leq \theta \leq 1.$$

As shown in class,

$$f(\theta|y) = \frac{\theta^y (1-\theta)^{n-y} \cdot g(\theta)}{\int_{-\infty}^{\infty} \theta^y (1-\theta)^{n-y} \cdot g(\theta) d\theta}.$$

Thus,

$$f(\theta|y) \propto \theta^y (1-\theta)^{n-y} \theta^{a-1} (1-\theta)^{b-1} = \theta^{y+a-1} (1-\theta)^{n-y+b-1}$$

from which we conclude that the posterior distribution of θ given y is $\beta(y+a, n-y+b)$.

4. (a) If Y_1, \dots, Y_n are i.i.d. $\text{Poisson}(\theta)$ random variables, then

$$f(y_1, \dots, y_n | \theta) = \prod_{i=1}^n f(y_i | \theta) = \frac{1}{\prod y_i!} e^{-n\theta} \theta^{\sum y_i}$$

where the sum and product both run from $i = 1$ to n .

4. (b) If the prior distribution of θ is $\Gamma(\alpha, \beta)$, then

$$f(\theta | y_1, \dots, y_n) \propto e^{-n\theta} \theta^{\sum y_i} \cdot \theta^{\alpha-1} e^{-\theta/\beta} = e^{-\theta(n+1/\beta)} \theta^{\alpha-1+\sum y_i}$$

which implies that the posterior distribution of θ given (y_1, \dots, y_n) is

$$\Gamma\left(\alpha + \sum_{i=1}^n y_i, \frac{1}{n + 1/\beta}\right).$$

5. (a) As shown in class, if Y_1, \dots, Y_n are a random sample of $\mathcal{N}(\theta, \sigma^2)$ random variables with prior density $g(\theta) \sim \mathcal{N}(\mu, \tau^2)$, then the posterior of θ given (y_1, \dots, y_n) is

$$\mathcal{N}\left(\frac{\sigma^2 \mu + n\tau^2 \bar{y}}{\sigma^2 + n\tau^2}, \frac{\sigma^2 \tau^2}{\sigma^2 + n\tau^2}\right).$$

Thus, using the data in the problem, we have

$$\sigma^2 = 20^2, \quad \mu = 180, \quad \tau^2 = 40^2$$

which implies that the posterior of θ given $\bar{y} = 150$ is

$$\begin{aligned} \mathcal{N}\left(\frac{20^2 \cdot 180 + n40^2 \cdot 150}{20^2 + n40^2}, \frac{20^2 \cdot 40^2}{20^2 + n40^2}\right) &= \mathcal{N}\left(\frac{72000 + 240000n}{400 + 1600n}, \frac{640000}{400 + 1600n}\right) \\ &= \mathcal{N}\left(\frac{180 + 600n}{1 + 4n}, \frac{1600}{1 + 4n}\right). \end{aligned}$$

5. (b) As shown in class, the posterior predictive distribution for \tilde{y} given y is

$$\mathcal{N}(\nu, \sigma^2 + \phi^2)$$

where

$$\nu = \frac{\sigma^2 \mu + n\tau^2 \bar{y}}{\sigma^2 + n\tau^2} \quad \text{and} \quad \phi^2 = \frac{\sigma^2 \tau^2}{\sigma^2 + n\tau^2}.$$

Thus, since $\sigma^2 = 20^2$, we find

$$f(\tilde{y} | y) \sim \mathcal{N}\left(\frac{180 + 600n}{1 + 4n}, 400 + \frac{1600}{1 + 4n}\right).$$