Bayesian Analysis Assignment One

Terry Liu 刘骏杰 1630005038

Question One Answer: For $\sigma = 2$, we can notice that:

$$p(y) = Pr(\theta = 1)p(y|\theta = 1) + Pr(\theta = 2)p(y|\theta = 2)$$
$$= 0.5N(y|1, 2^{2}) + 0.5N(y|2, 2^{2})$$

The graph can be drawn by R:

```
y <- seq(-7,10,.02)
density <- 0.5*dnorm(y,1,2) + 0.5*dnorm(y,2,2)
plot (y, density, ylim=c(0,1.1*max(density)))</pre>
```

The plot is:

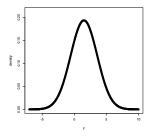


图: Marginal probability density of y

(b)

$$\begin{split} Pr(\theta=1|y=1) &= \frac{p(\theta=1,\ y=1)}{p(\theta=1,\ y=1) + p(\theta=2,\ y=1)} \\ &= \frac{Pr(\theta=1)p(y=1|\theta=1)}{Pr(\theta=1)p(y=1|\theta=1) + Pr(\theta=2)p(y-1|\theta=2)} \\ &= \frac{0.5N(1|1,2^2)}{0.5N(1|1,2^2) + 0.5N(1|2,2^2)} \approx 0.534 \end{split}$$

(c) If $\sigma \to$, the posterior density for θ will approaches to the prior density and the probability will be $\frac{1}{2}$; if $\sigma \to 0$ the $0.5N(1|2,^2) \approx 0$ so the $Pr(\beta = 1|y = 1) \to 1$.

Question Two Answer: From the question, we can notice that: $Pr(fraternaltwins) = \frac{1}{125}$ and $Pr(identicaltwins) = \frac{1}{300}$. Since the fraternal twins are in different perm cell. the $Pr(bothboys) = \frac{1}{4}$

The conditional probability that Elvis was an identical twins is :

$$\begin{split} Pr(identical twins | twins brother) &= \frac{Pr(identical twins,\ twins brother)}{Pr(twins brother)} \\ &= \frac{\frac{1}{2} \times \frac{1}{300}}{\frac{1}{2} \times \frac{1}{300} + \frac{1}{4} \times \frac{1}{125}} \\ &= \frac{5}{11} \approx 45.5\% \end{split}$$

Question Three Answer: We can just classify the genes: Xx, xX in one group and let Xx represent the group. Pr(child is $Xx \mid$ child has brown eyes

, the parents have brown eyes)

$$= \frac{0 \times (1-p)^4 + \frac{1}{4} \times (1-p)^3 + \frac{1}{2} \times 4p^2(1-p)^2}{1 \times (1-p)^4 + 1 \times 4p(1-p)^3 + \frac{3}{4} \times 4p^2(1-p)^2}$$

$$= \frac{2p(1-p) + 2p^2}{(1-p)^2 + 4p(1-p) + 3p^2}$$

$$= \frac{2p}{1+2p}$$

This is the poterior probability, but it is also the prior probability of the **probability that Judy is a heterozygote**, also, there added condition is that: **her** n **children are brown-eyed** $\Rightarrow Xx$ genes.

 $\Pr(\text{Judy is } Xx \mid n \text{ children all have brown eyes }, \text{ all previous information}) = \frac{\frac{2p}{1+2p} \times (\frac{3}{4})^n}{\frac{2p}{1+2p} \times (\frac{3}{4})^n + \frac{1}{1+2p} \times 1}}$ Also, we know that the Judy's Children has no blue-eyed, if we want Judy's

Also, we know that the Judy's Children has no blue-eyed, if we want Judy's grand that means Judy needs to be a heterozygote or Judy's child is Xx and Judy is XX: $Pr(Judy's child is <math>Xx \mid all information) = (Judy is <math>Xx \mid all information) + (Judy's child is <math>Xx \mid all information)$

$$\frac{\frac{2p}{1+2p} \times (\frac{3}{4})^n}{\frac{2p}{1+2p} \times (\frac{3}{4})^n + \frac{1}{1+2p}} \times \frac{2}{3} + \frac{\frac{1}{1+2p}}{\frac{2p}{1+2p} \times \frac{3}{4}^n + \frac{1}{1+2p}} \times \frac{1}{2}$$

Since her child is Xx, the probability of her grandchild having blue eyes is $0, \frac{1}{4}$ and $\frac{1}{2}$ as Judy's Child is XX,Xx or xx. so Pr(grandchild is xx|all the given information)

$$= \frac{\frac{2}{3} \frac{2p}{1+2p} (\frac{3}{4})^n + \frac{1}{2} \frac{1}{1+2p}}{\frac{2p}{1+2p} (\frac{3}{4})^n + \frac{1}{1+2p}} (\frac{1}{4} 2p(1-p) + \frac{1}{2} p^2)$$

$$= \frac{\frac{2}{3} \frac{2p}{1+2p} (\frac{3}{4})^n + \frac{1}{2} \frac{1}{1+2p}}{\frac{2p}{1+2p} (\frac{3}{4})^n + \frac{1}{1+2p}} (\frac{1}{2} p)$$

Quesion Four Answer: (a) The prior predictive distribution for y:

$$Pr(y = k) = \int_0^1 Pr(y = k|\theta)d\theta$$
$$= \int_0^1 \binom{n}{k} \theta^k (1 - \theta)^{n-k} d\theta$$
$$= \binom{n}{k} \frac{\Gamma(k+1)\Gamma(n-k+1)}{\Gamma(n+2)} = \frac{1}{n+1}$$

(b) We can notice that the posterior mean is $\frac{\alpha+y}{\alpha+\beta+n}$. We assume that $\lambda \in (0,1)$ and $\frac{\alpha+y}{\alpha+\beta+n} = \lambda \frac{\alpha}{\alpha+\beta} + (1-\lambda) \frac{y}{n}$

$$\frac{\alpha + y}{\alpha + \beta + n} = \lambda \frac{\alpha}{\alpha + \beta} + (1 - \lambda) \frac{y}{n}$$
$$\frac{\alpha + y}{\alpha + \beta + n} - \frac{y}{n} = \lambda (\frac{\alpha}{\alpha + \beta} - \frac{y}{n})$$
$$\lambda = \frac{\alpha + \beta}{\alpha + \beta + n}$$

(c) When we consider the uniform distribution, we can notice that $\alpha = \beta = 1$, so the prior variance is $\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)} = \frac{1}{12}$. In this situation, we can calculate the posterior variance:

Posterior Variance =
$$\frac{(1+y)(1+n-y)}{(2+n)^2(3+n)}$$
$$= (\frac{1+y}{2+n})(\frac{1+n-y}{2+n})(\frac{1}{3+n})$$

The sum of $\frac{1+y}{2+n}$ and $\frac{1+n-y}{2+n}$ is not greater than $\frac{1}{4}$. And $n \ge 1$, so $\frac{1}{3+n}$ is less than $\frac{1}{4}$. So Posterior vairance $\ne \frac{1}{16}$

(d) If $n=2,y=1,\alpha=1$ and $\beta=1$ The prior valance = $\frac{1}{12}$ and the posterior variance is 0.025